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THIS IS TO CERTIFY THAT THE THESIS PREPARED UNDER MY SUPERVISION BY

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ENTITLED A Parallel Equation-Based Algorithm for Simulating

Single Separation Columns

IS APPROVED BY ME AS FULFILLING THIS PART OF THE REQUIREMENTS FOR THE

DEGREE OF Bachelor of Science in Chemical Engineering

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**A Parallel Equation-Based Algorithm for Simulating  
Single Separation Columns**

**By**

**Daniel J. Kaiser**

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**Thesis**

**for the  
Degree of Bachelor of Science  
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## **ABSTRACT**

**A parallel equation-based algorithm has been developed to simulate single separation units. This algorithm was coded in FORTRAN and optimized for use on the Cray-2 Supercomputer, which has four processors available for multitasking. The algorithm was tested on several problems, and was generally reliable in converging to the correct solution. From the test results, the parallel program achieved speedups in excess of a factor of three over single processor operation. In addition, it was found that this algorithm, even on one processor, is inherently more efficient than normal Gauss-Jordan elimination. This new parallel algorithm clearly demonstrates that it can offer a much faster means for simulating distillation and absorber columns than traditional methods.**

## **ACKNOWLEDGEMENTS**

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## **I. INTRODUCTION**

Process flowsheeting, the use of computers to design and simulate chemical processes, has found extensive use in industry recently, as many commercial software packages have become available. These packages save the engineer time and also reduce human error by performing many of the tedious calculations previously done by hand. Also, flowsheeting packages allow the design engineer to "experiment" with a process by simply changing a design variable and letting the computer recalculate all the other variables. Of course, the faster the software can run, the easier it is for the engineer to achieve these goals, and speed depends on the computer technology available, which fortunately is evolving new architectures that are constantly increasing in both clockspeed and memory. As computer architectures evolve, new algorithms must be developed to exploit their increasing capabilities.

With mainframe computers, new technologies have resulted in two important advances: (1) vector processors, and (2) multi-processor machines. These powerful architectures, if properly used, can significantly decrease the run-time of many programs. As of now, these architectures are currently found only on the most advanced mainframes, such as the Cray supercomputers. In time, though, these types of computers will become less expensive and gain wider acceptance, finding their way into the mainstream of commercial computers. This will result in the need for a new generation of software to take advantage of vector and multi-processor machines. As computers evolve, so must the software.

This study focuses specifically on developing a new computer algorithm for the optimization of single separation units, distillation and

absorber columns, by using an equation-based method founded on the Newton-Raphson technique. However, what differentiates this algorithm from others is its attempt to utilize the advantages of a multi-processor machine. The goal therefore becomes constructing the program in such a way that a large portion of it can be "divided up" among the various processors, allowing them to work concurrently. Theoretically, the maximum speedup that can be achieved is a factor of  $n$ , where  $n$  is the number of processors available. This study determines how close one can approach this theoretical limit with a distillation algorithm written for the Cray-2 supercomputer.



## **II. ADVANCED COMPUTER ARCHITECTURES**

In most computers, the processor handles instructions one part at a time. For example, consider the following DO loop:

```
do 10 i = 1,100
  A(i) = B(i) + C(i)
10  continue
```

A normal processor will execute this loop in the following manner:

1. Fetch B(1) from storage
2. Fetch C(1) from storage
3. Add B(1) and C(1)
4. Store the result in A(1)
5. Increment i
6. Repeat

On the other hand, a vector processor, such as in the Cray X-MP, can do the same task in a much more efficient manner:

1. Fetch all of vector B from storage
2. At the same time, fetch vector C from storage
3. As soon as the first elements of both vectors arrive at the processor, add them together
4. As soon as addition is complete, store the result in A
5. Continue until the loop is completed

This process is illustrated in Figure 1:

Fortunately, the FORTRAN compilers on the Cray are capable of identifying and vectorizing such code; no special directives are required from the programmer, generally. However, the programmer must take care to avoid writing a portion of code that cannot vectorize. According to Levesque and Williamson (pp. 90-91), things to avoid include:

1. Recursion
2. Subroutine calls
3. References to external functions unknown to FORTRAN 77
4. Input/output statements
5. Assigned GOTO statements
6. Some nested IF blocks

7. GOTO statements that exit the loop
8. Backward transfers within a loop

If a programmer keeps these few items in mind while programming, he or she can easily develop code that the compiler can vectorize.

Another advanced architecture, with which this study is primarily concerned, is multi-processor machines. Like vector processing, it provides a means for completing several tasks simultaneously. However it differs in the granularity (or level) with which it does tasks simultaneously. For example, consider the following DO loop:

```

do 20 i = 1,256
    A(i) = B(i) + C(i)
20  continue

```

If there are four processors available, each can be assigned 64 iterations of the loop, and since each iteration is completely independent of the others, the processors can operate concurrently. And, of course, each processor can also be carrying on vector processing as well. Parallelism on the DO loop level, such as this, is known as microtasking (Cray Multitasking p. 4-1). Parallelism can occur at other levels as well. Consider the following piece of code:

```

call task1(arguments)
call task2(arguments)
call task3(arguments)
call task4(arguments)

```

Again if there are four processors available, and each subroutine is independent of the others, then these four subroutines can be run concurrently. Parallelism on the subroutine level, such as this, is known as macrotasking (Cray Multitasking p. 5-1).

Exploiting the use of several processors simultaneously requires much more programmer attention than does vector processing. Essentially, it is up to the programmer to develop an algorithm that will parallelize efficiently. Then once the algorithm is coded, special compiler directives or subroutine calls must be inserted to coordinate the operation of the processors. Currently, Cray is working on developing compilers that can recognize parallel code and automatically insert the correct directives (known as autotasking). However, these compilers are still in their developmental stages, and even when fully mature, do not eliminate the programmer's responsibility to develop code that can efficiently parallelize.

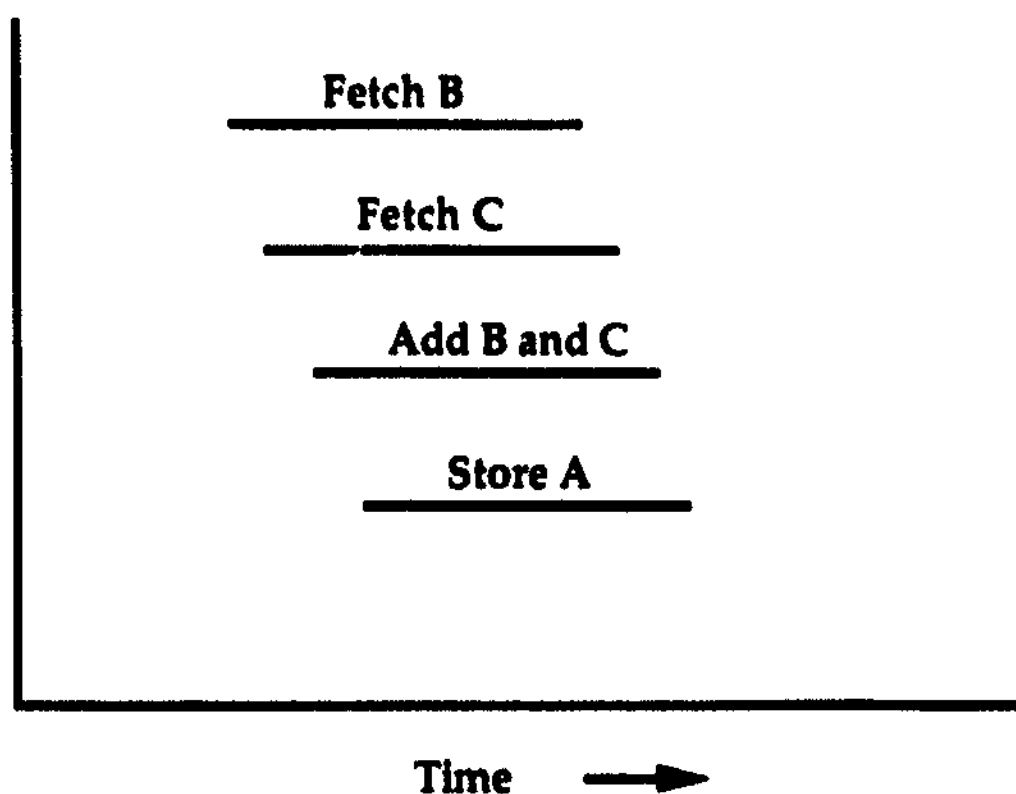


Figure 1

Timing Diagram of Vectorized Loop

(Levesque and Williamson p. 39-42)

### III. PARALLEL EQUATION-BASED METHOD

The method used here to optimize separation units is the equation-based method, whereby a set of non-linear equations are solved simultaneously. The advantage of this method over the sequential modular approach, a step-by-step iterative technique, is its speed. However, it requires significantly more computer memory, but with today's advanced mainframes, such as the Cray, memory size is increasingly becoming an insignificant concern.

For separation problems, the equations are obtained by doing mass, energy, and equilibrium balances around each tray. This is the formulation of Naphtali and Sandholm (1971) and is shown below:

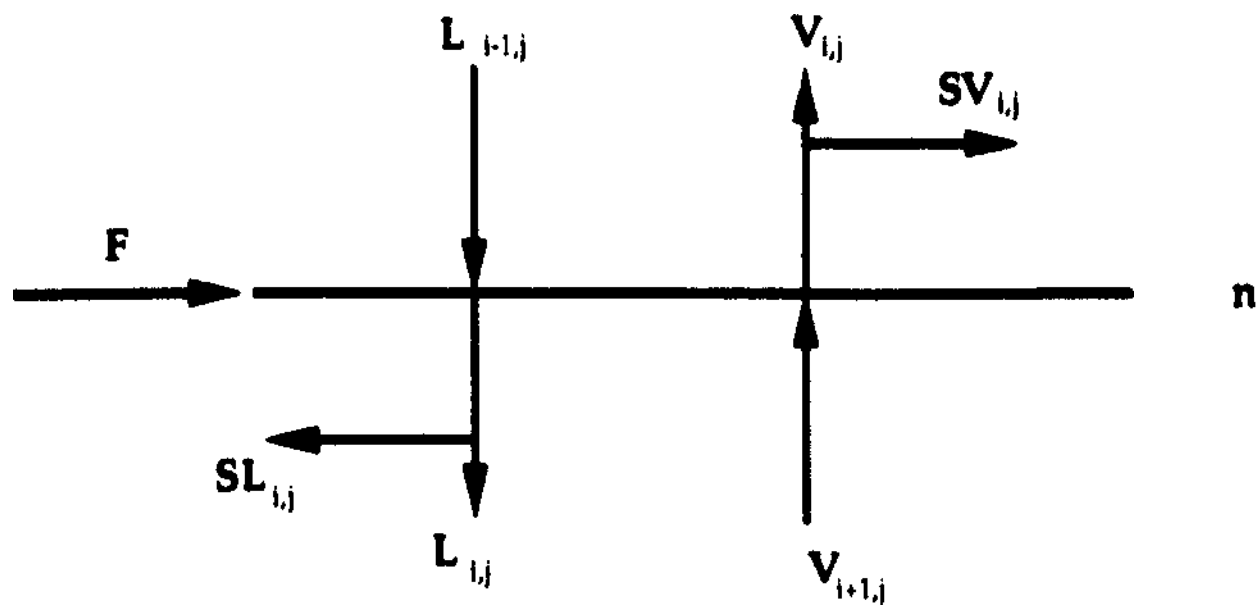


Figure 2  
Tray Configuration

The equations that result from each tray are as follows:

Mass:  $M_{i,j} = V_{i,j} + SV_{i,j} + L_{i,j} + SL_{i,j} - V_{i+1,j} - L_{i-1,j} - F_{i,j}$

Energy:  $E_i = HV_{i,j} + HSV_{i,j} + HL_{i,j} + HSL_{i,j} - HV_{i+1,j} - HL_{i-1,j} - HF_{i,j}$

Equilibrium:  $Q_{i,j} = n_i K_{ij} V_i L_{ij} / L_i - V_{ij} + (1-n_i) V_{i+1,j} V_i / V_{i+1}$

V = vapor flowrate

L = liquid flowrate

SL = liquid sidestream

SV = vapor sidestream

i = tray #

j = component #

K = equilibrium (y/x)

n = tray efficiency

The equations are set up as discrepancy functions, so that as the values of the variables approach the optimum, the functions will approach zero. Also, one should note that for each tray, there are j mass equations, j equilibrium equations, and one energy equation, for a total of  $2j + 1$  equations, and thus variables, per tray. Therefore, for the entire problem, there are  $(2j+1) \times (\text{\# of trays})$  variables, a number known as the *order* of the problem. This formulation provides an extremely rigorous simulation of the column and allows for multiple feeds and sidestreams. For absorbers, these equations can be used for all trays, but for distillation columns, the energy equation for both the condenser and reboiler must be substituted with one of the following:

- |                        |  |
|------------------------|--|
| 1. specify Q           | merely add Q to energy balance                 |
| 2. reflux/reboil ratio | $L = V(R)$ or $V = L(R)$                       |
| 3. specify temp        | $T = T_{\text{spec}}$                          |
| 4. specify flowrate    | $V = V_{\text{spec}}$ or $L = L_{\text{spec}}$ |

These substitutions are easy to incorporate and provide the user with several options from which to choose.

To solve all of these equations, the algorithm employs the Newton-Raphson method:

$$(1) \quad \Delta x^{m+1} = - (J^{-1} |_{x^m}) F^m$$

where  $x^m$  is the vector of variables after the  $m^{\text{th}}$  iteration,  $F^m$  is the vector of discrepancy functions evaluated at  $x^m$ ,  $J^{-1}$  is the Jacobian matrix ( $dF/dx$ ), and  $\Delta x^{m+1}$  is the update to the variables:

$$(2) \quad x^{m+1} = x^m + f \Delta x^{m+1}$$

and  $f$  is a factor used to select the optimal fraction of the update vector.

Now, to evaluate the Jacobian, the algorithm must calculate a host of derivatives. Fortunately, the set of equations for each tray contains only variables from itself and the trays above and below it, which means that most of the elements of the Jacobian are zeroes. Overall, the Jacobian is constructed as shown in Figure 3. Actually, each element in the matrix is a *block* of elements. For example, the upper left element can be expanded as shown in Figure 4:

$$J = \begin{bmatrix} \frac{dF_1}{dX_1} & \frac{dF_1}{dX_2} & \cdots & \frac{dF_1}{dX_n} \\ \frac{dF_2}{dX_1} & \frac{dF_2}{dX_2} & \cdots & \frac{dF_2}{dX_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{dF_n}{dX_1} & \frac{dF_n}{dX_2} & \cdots & \frac{dF_n}{dX_n} \end{bmatrix}$$

Figure 3  
Jacobian Matrix

$$\frac{dF_1}{dX_1} =$$

	$V_{11}$	$V_{12}$	$\dots$	$V_{1k}$	$T_1$	$L_{11}$	$L_{12}$	$\dots$	$L_{1k}$
$M_{11}$	$\frac{dM_{11}}{dV_{11}}$	$\frac{dM_{11}}{dV_{12}}$	$\dots$	$\frac{dM_{11}}{dV_{1k}}$	$\frac{dM_{11}}{dT_1}$	$\frac{dM_{11}}{dL_{11}}$	$\frac{dM_{11}}{dL_{12}}$	$\dots$	$\frac{dM_{11}}{dL_{1k}}$
$M_{12}$	$\frac{dM_{12}}{dV_{11}}$	$\dots$		$\frac{dM_{12}}{dV_{1k}}$	$\vdots$	$\frac{dM_{12}}{dL_{11}}$	$\dots$		$\frac{dM_{12}}{dL_{1k}}$
$\vdots$	$\vdots$			$\vdots$	$\vdots$	$\vdots$			$\vdots$
$M_k$	$\frac{dM_k}{dV_{11}}$	$\dots$		$\frac{dM_k}{dV_{1k}}$	$\frac{dM_k}{dT_1}$	$\frac{dM_k}{dL_{11}}$	$\dots$		$\frac{dM_k}{dL_{1k}}$
$E_1$	$\frac{dE_1}{dV_{11}}$	$\dots$		$\frac{dE_1}{dV_{1k}}$	$\frac{dE_1}{dT_1}$	$\frac{dE_1}{dL_{11}}$	$\dots$		$\frac{dE_1}{dL_{1k}}$
$Q_{11}$	$\frac{dQ_{11}}{dV_{11}}$	$\frac{dQ_{11}}{dV_{12}}$	$\dots$	$\frac{dQ_{11}}{dV_{1k}}$	$\frac{dQ_{11}}{dT_1}$	$\frac{dQ_{11}}{dL_{11}}$	$\frac{dQ_{11}}{dL_{12}}$	$\dots$	$\frac{dQ_{11}}{dL_{1k}}$
$Q_{12}$	$\frac{dQ_{12}}{dV_{11}}$	$\dots$		$\frac{dQ_{12}}{dV_{1k}}$	$\vdots$	$\frac{dQ_{12}}{dL_{11}}$	$\dots$		$\frac{dQ_{12}}{dL_{1k}}$
$\vdots$	$\vdots$			$\vdots$	$\vdots$	$\vdots$			$\vdots$
$Q_k$	$\frac{dQ_k}{dV_{11}}$	$\dots$		$\frac{dQ_k}{dV_{1k}}$	$\frac{dQ_k}{dT_1}$	$\frac{dQ_k}{dL_{11}}$	$\dots$		$\frac{dQ_k}{dL_{1k}}$

Figure 4  
Block Element of Jacobian

Evaluating a matrix of this size could be extremely expensive computationally. But as mentioned earlier, a vast majority of the blocks are actually zeroes. In fact, the Jacobian matrix takes on a very symmetric form, known as the block-tridiagonal form. It is this very fact that allows distillation problems to be solved in parallel, and how this is done will be discussed later. Figure 5 below shows the overall structure of the Jacobian once it has been evaluated:

$$\begin{array}{ccccccccc}
 B_1 & C_1 & 0 & 0 & 0 & \cdots & & \cdots & 0 \\
 A_2 & B_2 & C_2 & 0 & 0 & \cdots & & \cdots & 0 \\
 0 & A_3 & B_3 & C_3 & 0 & \cdots & & \cdots & 0 \\
 \vdots & & & & & & & & \vdots \\
 & & & & & & & & \\
 & & & & & & 0 & A_{n-2} & B_{n-2} & C_{n-2} & 0 \\
 & & & & & & 0 & 0 & A_{n-1} & B_{n-1} & C_{n-1} \\
 0 & 0 & 0 & \cdots & & & 0 & 0 & 0 & A_n & B_n
 \end{array}$$

Figure 5  
Evaluated Jacobian

Once the Jacobian has been evaluated, rearranging equation (1) and solving using Gauss-Jordan elimination will yield the values for the variable update vector  $\Delta x$ .

If one were to write a program to implement this entire method, one would find that it spends a vast majority of its time working on the Gauss-Jordan elimination step. Since this step is the main bottleneck in the program, if one can speed it up, then one can effectively speed up the entire program. To accomplish this, an algorithm developed by Sameh (p. 37-57) is



utilized to perform the Gauss-Jordan step in parallel. To illustrate how this algorithm works, say for example the computer has four processors available. The algorithm then partitions the Jacobian (Figure 5) into four different sections, each one the responsibility of a unique processor. The processors simultaneously perform Gauss-Jordan elimination, each operating on its own section, to produce the partially solved matrix shown in Figure 6 (note the asterisks represent fill-in which occurs during the Gauss-Jordan process and is why Figure 6 cannot be immediately solved for the answers):

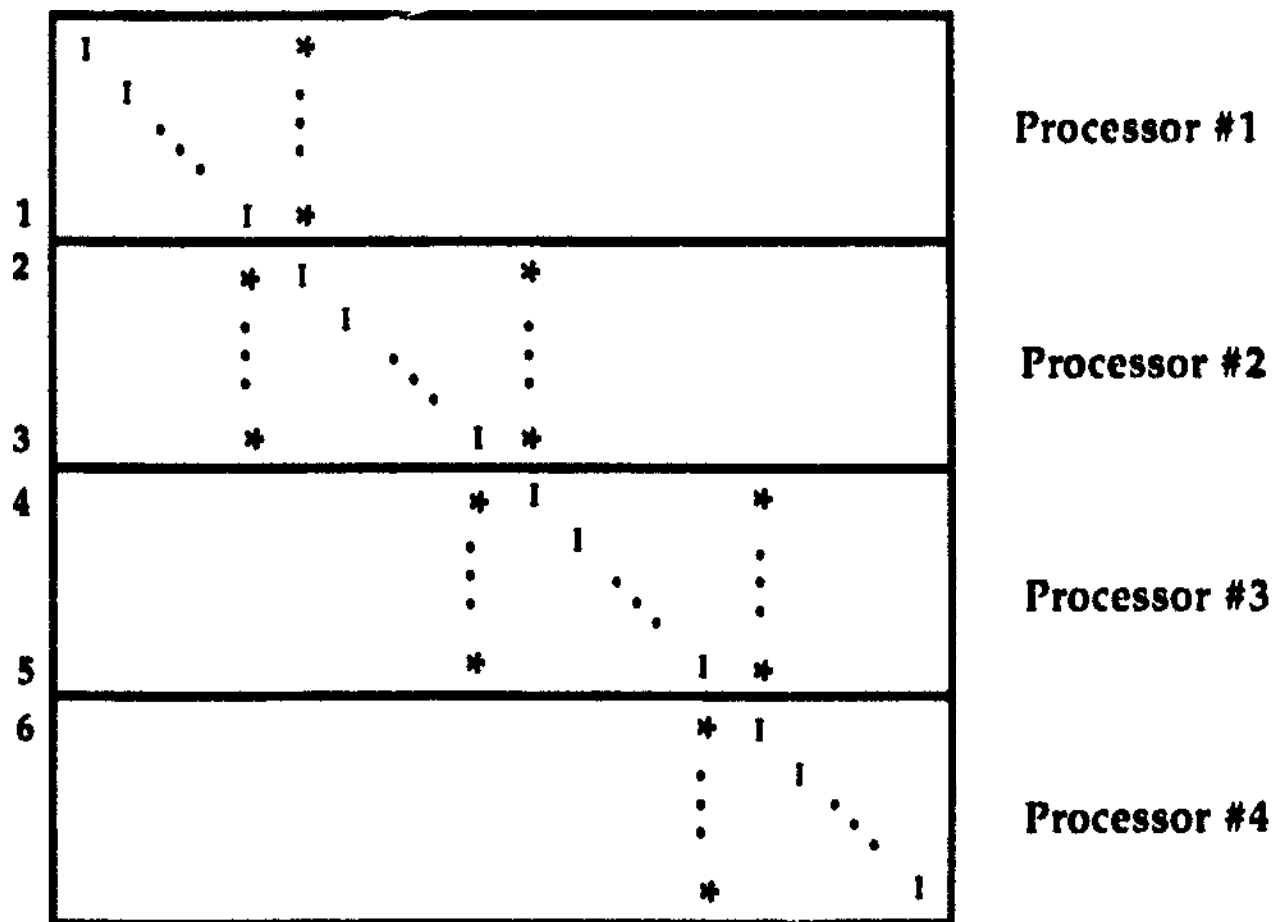


Figure 6  
Results of Parallel Gaussian Elimination

Because of the fill-in that results, the system still cannot be solved. However, if one merely selects the rows of the matrix that border the processor partitions (numbered 1 through 6 in Figure 6) and place these rows in another matrix, they form a complete set of equations that can be solved explicitly (See Figure 7). Once the values for these variables are known, the remaining variables can be calculated with back-substitution on the matrix in Figure 6.

At this point one might be tempted to conclude that as the number of available processors increases, the execution time will increase. However this may be true only up to a certain point because as the number of equations increases, the size of the matrix in Figure 7 will also increase, and this matrix cannot be solved efficiently in parallel. As a result, this algorithm would seem best suited to run on a machine with a few very powerful processors rather than one with many less powerful processors. The Cray-2 fits the former category very well as it has four extremely fast processors (4-nanosecond clock cycle) and so will be the machine used in this study.

	1	2	3	4	5	6
1	I	*	0	0	0	0
2	*	I	0	*	0	0
3	*	0	I	*	0	0
4	0	0	*	I	0	*
5	0	0	*	0	I	*
6	0	0	0	0	*	I

Figure 7

Extracted Matrix

#### IV. COMPUTER PROGRAM

The computer program to implement this parallel algorithm is shown in Appendix A, but some discussion of how it works is necessary before moving on. The program has been written to be, relatively speaking, somewhat user friendly, which means that a person with knowledge on how the input files should be constructed can configure the program to solve many different types of problems simply by adjusting the dimensions of the arrays and adjusting the values contained in the PARAMETER statements. Actually, the program could have gone one step further in friendliness by simply overdimensioning the arrays to handle all reasonable types of problems and by reading in the values currently defined in PARAMETER statements. However, this was not done because these features would cause the program to run more slowly. If the arrays are overdimensioned, a small-size problem will run slowly because the computer must shuffle around large portions of arrays that are not even being used. The values defined in PARAMETER statements are there because they are frequently used as counter limits in DO loops, and the Cray FORTRAN compiler is better able to vectorize a DO loop when its length is exactly known. In other words, some user friendliness is sacrificed in order to maximize the speed of the program, but not to an extent where it is too cumbersome to use.

The main program first calls subroutine INPUT which reads in all the necessary data, such as initial guesses, thermodynamic data, and user options. Next, the program calls DISCREP which evaluates the discrepancy functions (F) using the current values for the variable vector (x). Then the program calls DERIV which evaluates the Jacobian using the current variable values. When completed, the program then enters the Gauss-Jordan elimination

phase. Depending on what option the user selects, the program will either do Gauss-Jordan elimination in a normal procedure (REGSOLVE) or it will begin the Sameh algorithm. For the Sameh algorithm, the first step is a call to PARTIAL which is the subroutine that executes Gauss-Jordan elimination in parallel until the matrix in Figure 6 is obtained. Once this step is complete, the program calls SELECT which collects the lines of the Jacobian matrix reordering the processor partitions and condenses them in the matrix shown in Figure 7. A call to FSTSOLVE then quickly solves this short set of equations. Finally, SCATTER1 and SCATTER2 take the update variables calculated by FSTSOLVE and perform back-substitution to obtain the remaining update variables, completing the Sameh Gauss-Jordan elimination algorithm.

At this point, the program calls SEARCH which selects the best fractional portion of the update vector. Recalling equation (2):

$$(2) \quad x^{m+1} = x^m + f \Delta x^{m+1}$$

SEARCH determines the value for  $f$  (between 0 and 1) that will minimize the sum of the squares of the values of the discrepancy functions and then carries out equation (2) to calculate the new values for the variable vector. Next, the program calls REALVAR which checks to make sure that none of the new variables have passed into a physically meaningless region (such as a negative flowrate). If it finds such a value, it adjusts the value so that it no longer is unrealistic.

Now the program tests for convergence by first calling DISCREP to evaluate  $F$  with the new values for  $x$  and then SUM to determine the sum of the squares. If the value returned by SUM is less than some tolerance, or the program has completed a maximum number of iterations, the program exits

the iteration loop, prints out the final values (subroutine OUT), and quits. If another iteration is necessary, the program returns to the top for another run through the routines.

As mentioned before, there are two ways to implement parallel processing on the Cray-2--microtasking and macrotasking. This program utilizes microtasking and is implemented by inserting special compiler directives in the program. The program is then compiled using the following command:

**CF77 -Zm filename**

Inserting -Zm tells the compiler that microtasking compiler directives (CMIC\$) are to be interpreted. If -Zm is left off, the compiler merely treats the directives as comment statements (since they begin with a "c"). In the program itself, the first directive, CMIC\$ GETCPUS 4, appears in the main program very near the beginning, and instructs the computer to get all four CPUs focused on the program. However, even though all the processors have been "fetched" at this point, only one will be working on the program at any given time until PARTIAL is reached. The code for PARTIAL is preceded with the directive CMIC\$ MICRO which tells the computer that this is a microtasked subroutine. PARTIAL itself then consists of one giant DO loop which iterates four times. The first iteration does Gauss-Jordan elimination on the top quarter, the second iteration on the second quarter and so on. Therefore, each iteration of this loop is completely independent of the others, which means the loop is ideal for parallelization. Inserting the directive CMIC\$ DO GLOBAL just in front of the loop tells the computer this loop can be split up among the processors fetched by the GETCPUS directive.

Additionally, subroutine DERIV is also microtasked in much the same way since the evaluation of each tray is completely independent of each other. Implementing microtasking is actually not very difficult. All one needs to do is insert just a few compiler directive in the correct places. The compiler then takes care of the rest by interpreting these commands and actually inserting more lines into the program to make calls to the Cray microtasking functions and routines, which then guide the operation of the processors. Cray has made implementing microtasking easy, but they still have not taken away the responsibility of the programmer to construct code that will parallelize, which in this case means a DO loop with completely independent iterations. Microtasking can be easy, but the code must be properly written.

## V. RESULTS AND DISCUSSION

Once the program was written and debugged, testing began to determine the robustness of the program. First, the program was tested on an absorber problem that was previously used by Naphali and Sandholm. The details of the problem are shown in Appendix B, but essentially it is an absorber tower with 20 trays and four components with a wide range in volatility. Applying the formulas derived earlier, this makes for a total of

$$(2 \times 4 + 1) \times 20 = 180 \text{ variables}$$

The program was able to solve this problem in five iterations to a sum of the squares of less than  $1 \times 10^{-15}$ . The compositions of the tops and bottoms streams are shown in Appendix B as well. Next, to test the programs ability to handle distillation columns, a distillation problem, again from Naphali and Sandholm, was used. This problem featured three hydrocarbons with similar volatilities, and twenty trays, with the top and bottom trays being a partial condenser and partial reboiler respectively. Furthermore, a sidestream was drawn off the condenser. Again applying the formulas, this amounts to

$$(2 \times 3 + 1) \times 20 = 140 \text{ variables}$$

This problem converged in 4 iterations to a sum of the squares of less than  $1 \times 10^{-10}$ . At this point it was clear the program was capable of handling both absorbers and distillation columns, as well as sidestreams, and wide ranges in volatility.

The next step was to collect data on the efficiency of the algorithm itself by determining the following relationships:

1. wallclock time and MFLOPS versus number of variables using one CPU

2. wallclock time and MFLOPS versus number of variables using four CPUs
3. wallclock time and MFLOPS versus number of variables using normal Gauss-Jordan elimination
4. wallclock time versus block size

Wallclock time is defined as the amount of processor time needed to execute the program in a dedicated environment using just one processor. This provides a very direct way of evaluating the speed of the program. MFLOPS (millions of floating point operations per second) also indicates the efficiency of the program by tallying the amount of "work" the computer does for a unit of time. The higher the MFLOPS, the more efficient the program.

To collect this data, the absorber problem was used for all runs to maintain uniformity. The first three relationships were obtained by varying the number of trays (and thus the number of variables) of the absorber, while the fourth relationship was obtained by adding more components to the absorber (and thus increasing the block size).

Figure 8 shows the wallclock results and Figure 9 the MFLOPS for the original problem (block size 9). Comparing the results on one CPU and four CPUs shows that a significant amount of speedup occurred from microtasking. This is shown in Figure 10. Clearly, the speedup increases with the number of variables, which is due to the fact that as the number of variables increases, the program spends a greater and greater fraction of its time in PARTIAL and DERIV, which are the microtasked routines. At the highest value, speedups of about a factor of three occurred. Recall that the maximum theoretical speedup would be four, using four processors. In practice, a speedup of three is considered very good.



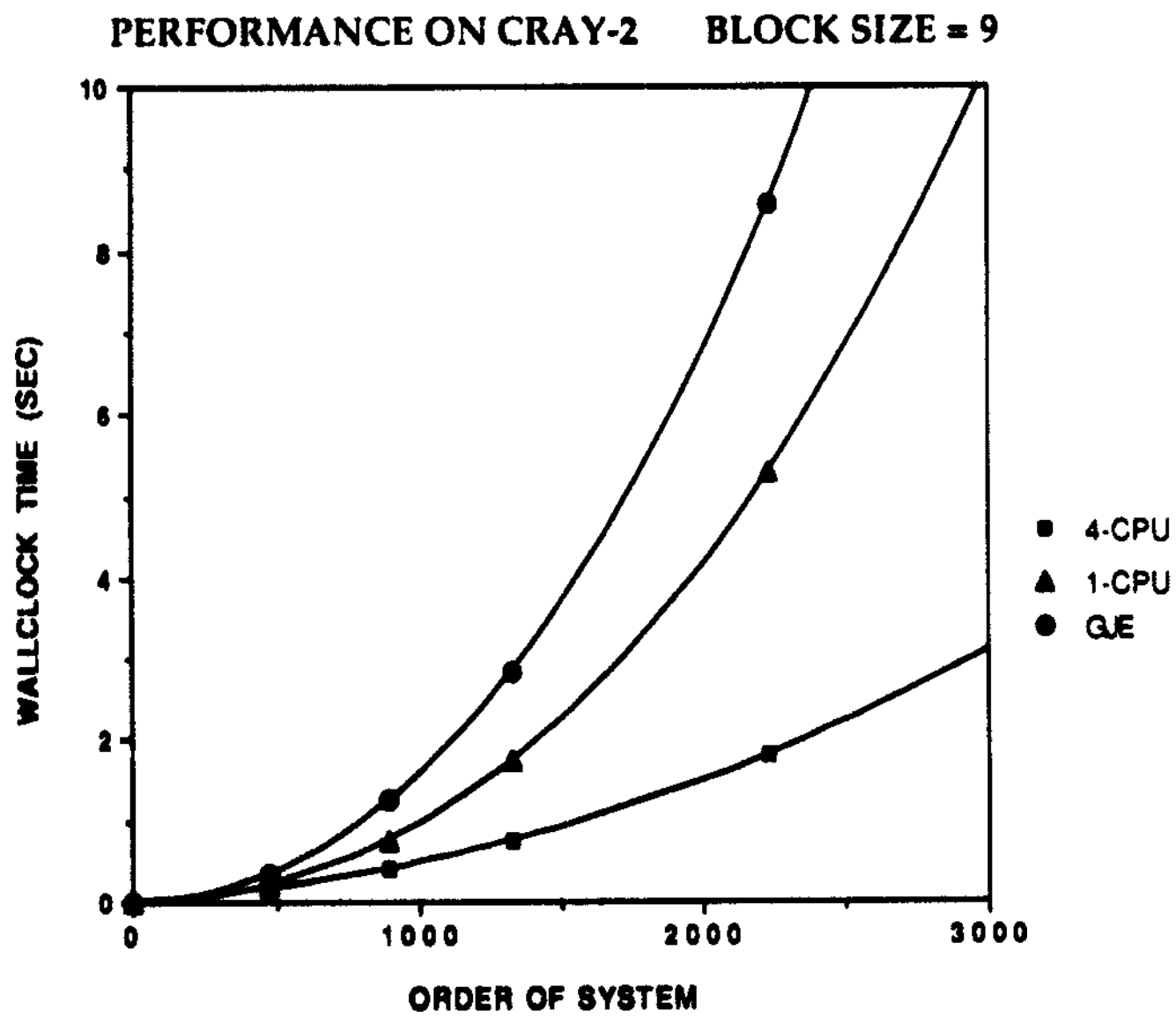


Figure 8  
Wallclock Results for Blocksize = 9

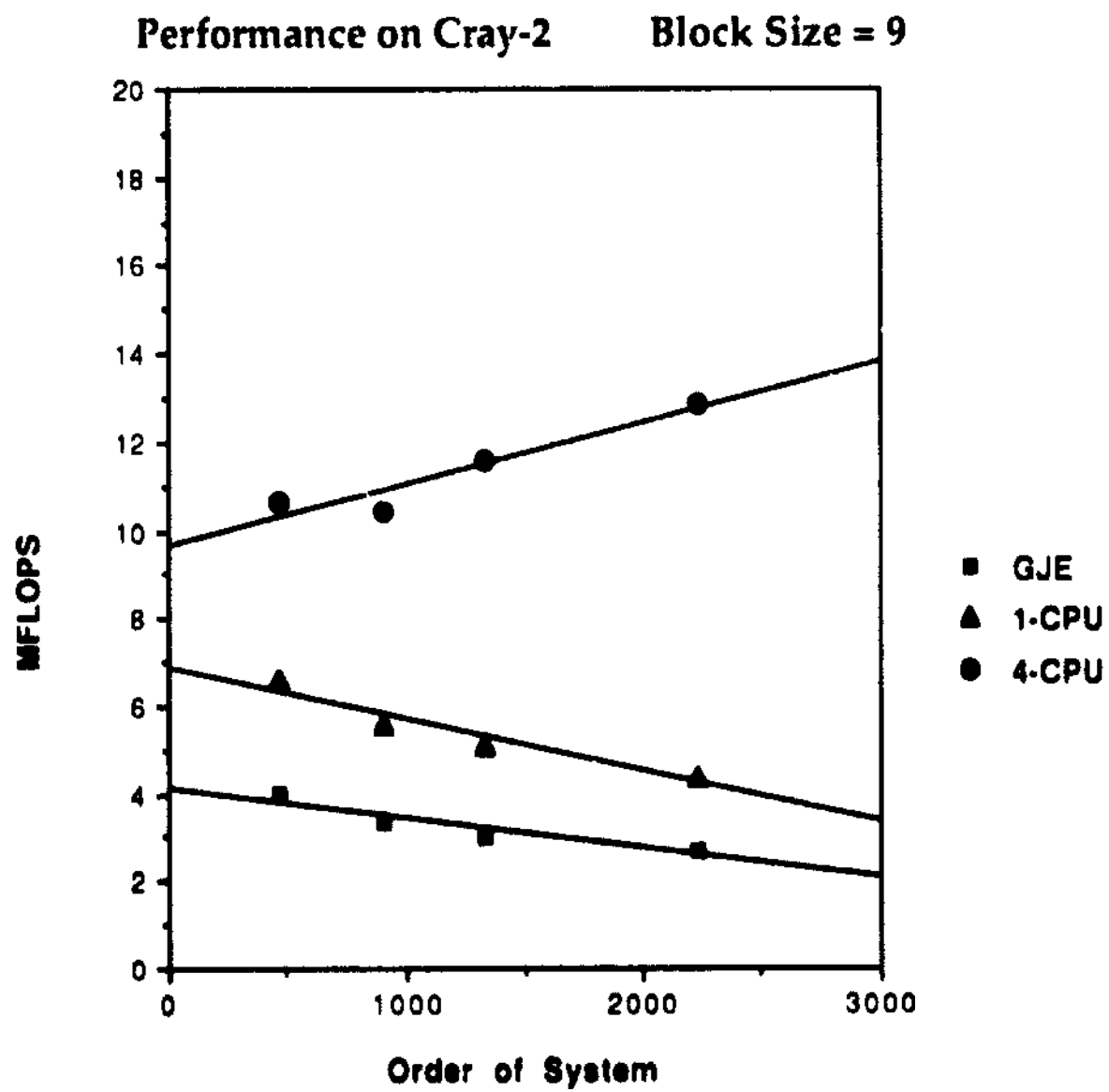


Figure 9  
MFLOPS for Blocksize = 9

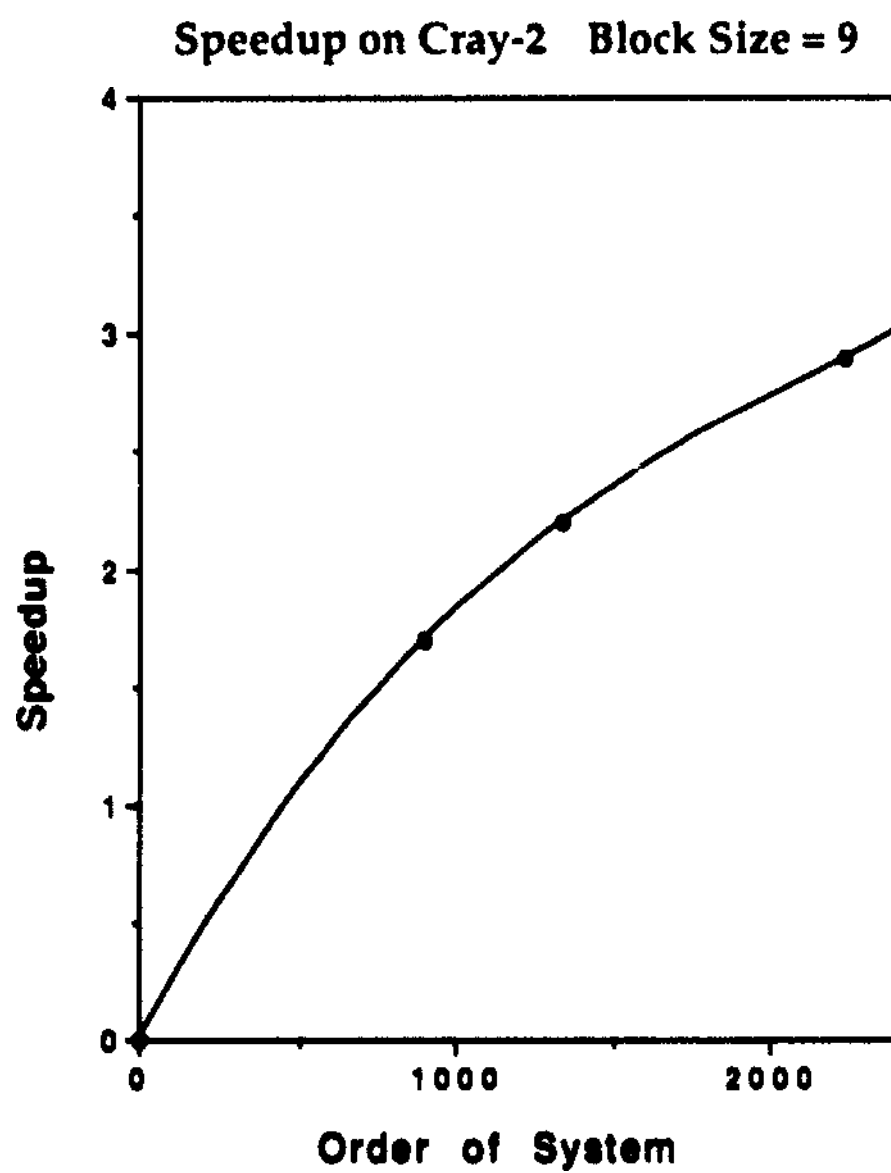


Figure 10  
Speedup versus Order for Blocksize = 9

Another result that is somewhat mystifying at first is that the Sameh algorithm on one CPU actually ran faster than did normal Gauss-Jordan elimination. This was not expected since the Sameh algorithm has additional tasks to do, such as solving the reduced matrix, that normal Gauss-Jordan elimination does not. However, a closer examination of the two algorithms reveals a fundamental difference. As the block solver moves down the diagonal, it pivots to produce zeroes everywhere above and below the diagonal in the column. Since the Jacobian is in block form, it need only go down to the bottom edge of the block, since everything below that is already zero. However, it must go all the way to the top of the entire matrix to produce zeroes since fill-in will have occurred during the operations performed on previous columns. As it proceeds down along the diagonal, by the time it reaches the bottom, it is pivoting all the way up the entire length of the matrix. On the other hand, with the Sameh algorithm, the farthest it must pivot would be up to the top of the section, which in this case would be no more than one quarter of the whole matrix. This fundamental difference results in many fewer calculations that the Sameh algorithm must execute to achieve the same results as the normal Gauss-Jordan algorithm. The Sameh algorithm, even on one processor, is a clearly more efficient algorithm. The speedup of the one CPU Sameh algorithm over the normal Gauss-Jordan algorithm is shown in Figure 11 and is essentially constant with the order of the system, always between 1.6 and 1.7.

Figure 12 shows wallclock times for a different block size (21). Again, the Sameh algorithm was successful in significantly speeding up the operation of the program, reaching speedups of greater than three for the largest orders (Figure 13). In fact, if one compares Figures 13 and 10, one will

see that for a given order of the system, the program achieves a greater speedup with the larger block size. However, the program requires a longer overall wallclock time with the larger size, outweighing the increased speedup. The net effect is that an increased block size results in increased speedup, but slower total execution time for a given order of the system.

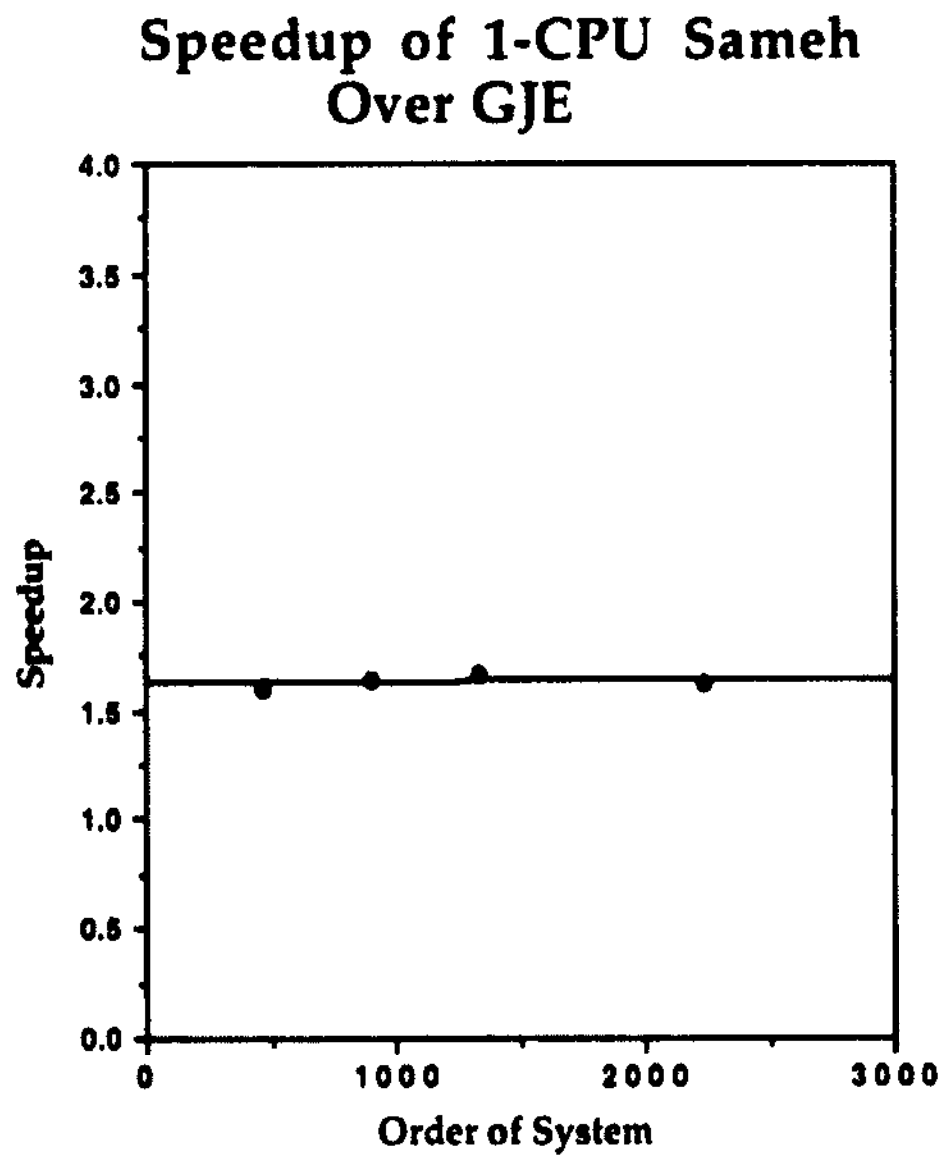


Figure 11

Speedup of Sameh Algorithm over GJE

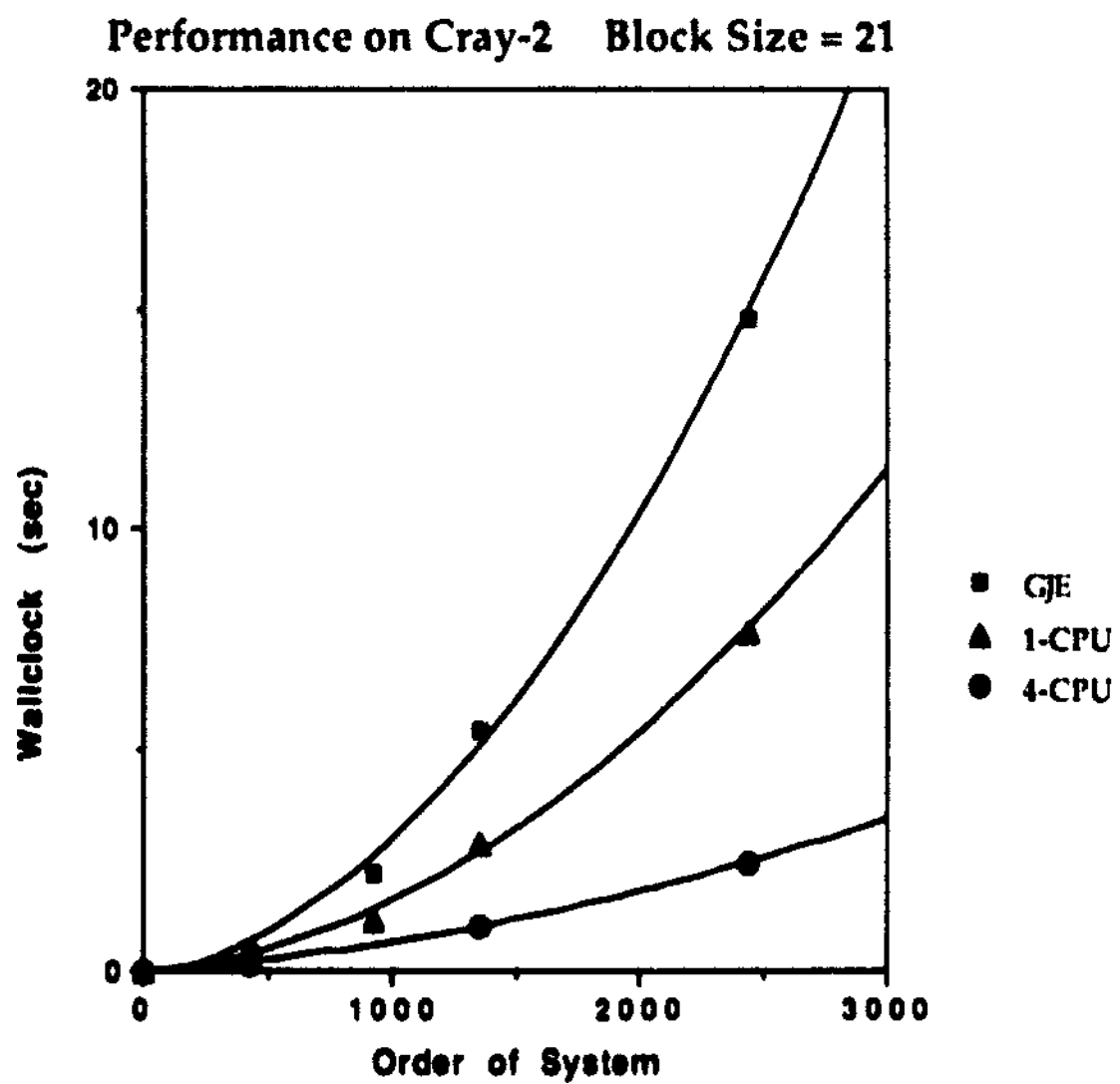


Figure 12  
Wallclock Results for Blocksize = 21

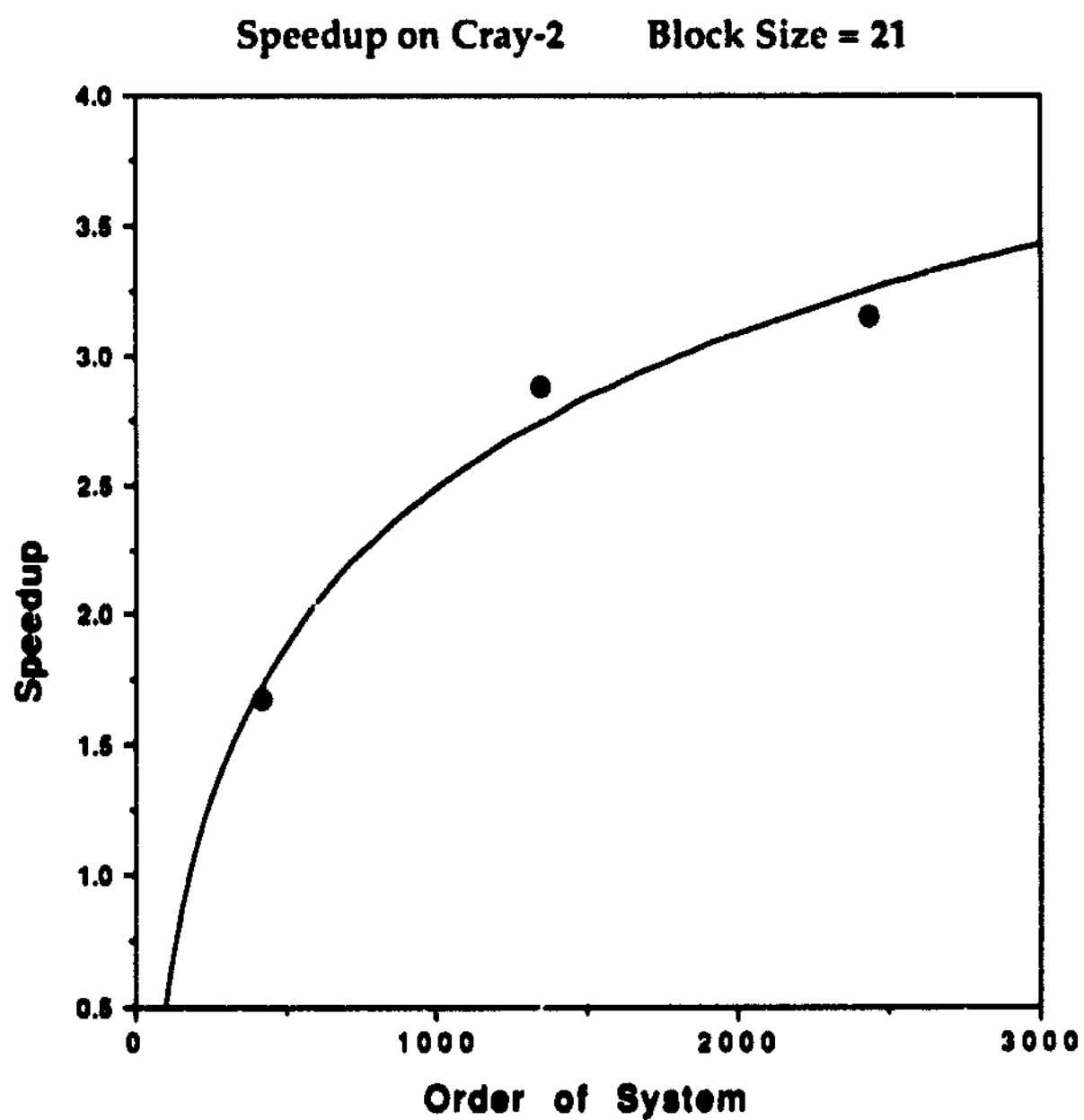


Figure 13  
Speedup versus Order for Blocksize = 21



## **CONCLUSIONS AND RECOMMENDATIONS**

The following conclusions can be drawn about the Sameh algorithm as applied to single separation columns:

- (1) The Sameh algorithm on four CPU's is capable of achieving speedups in excess of a factor of 3 over normal one CPU operation. Speedup increases with the order of the system and with the block size of the problem.
- (2) The Sameh algorithm on one CPU is actually more efficient than normal Gauss-Jordan elimination by a factor of about 1.7.
- (3) The Sameh algorithm is robust in that it can handle any number of trays, components, and variables, and can solve both distillation and absorber problems.

As for further study, the following recommendations can be made:

- (1) Further runs could be made with even greater system orders (5000 to 10000 variables) in order to determine the asymptotic speedup limit for a given blocksize.
- (2) Further attempts can be made to parallelize the remaining subroutines of the program in order to push the maximum speedup closer to 4.
- (3) The program can be run on other computer systems (such as the Alliant) which offer larger numbers of processors (64 or more). Runs varying

the number of processors available could determine at what point adding additional processors becomes counterproductive (the size of the reduced matrix becomes too large).

Clearly, the Sameh algorithm is an effective, robust method for solving single separation columns much more quickly than non-parallel algorithms. The next area of research for this topic now will be to attempt to apply this algorithm to systems of interlinked columns. Interlinked systems do not have a true block tri-diagonal structure due to the presence of off diagonal blocks where the columns are connected to each other. The presence of off diagonal blocks will effect the way in which the load must be divided among the available processors, and for some problems the load perhaps may not be efficiently distributed at all. It will be interesting to see if this algorithm can be adopted to handle these more complex systems, because if it can, it offers the potential for greatly increasing the speed of execution of these types of problems.

**APPENDIX A**

**COMPUTER PROGRAM**

C program SEPARATE is designed to simulate absorber and  
 C distillation columns and is optimized for the Cray-2  
 C  
 C author: Daniel J. Kaiser

program SEPARATE

```

real x(-4:473),c(468,-9:477),f(468),n(52),y(20),fd(468),
+ s(468),yv(4),bv(4),yl(4),bl(4),data(2),xnew(468),sums,tol
+ ,ctemp(30,30),ftemp(30),xtemp(30)

```

```

integer iter,mxiter,i,j,m,k,tray,cp,v,choice,cvar,part,dist(7)

```

C v is the number of variables  
 C cp is the number of components  
 C tray is the number of trays  
 C cvar is the size of the post-Sameh condensed matrix  
 C and is equal to  $(cp+1)*6$

C  $x(-cp,v+cp+1)$  is the vector of variable values  
 C  $c(v,-2*cp-1:v+2*cp+1)$  is the Jacobian matrix  
 C  $f(v)$  is the vector of discrepancy functions  
 C  $n(tray)$  is a vector containing tray efficiencies  
 C  $y(5*cp)$  is a vector of K values. The K-values are  
 C to be fitted to a 4th order polynomial:  
 C  $K = AT**4 + BT**3 + CT**2 + DT + E$   
 C  $y(1) = A$  for component 1  
 C  $y(2) = B$  for component 1  
 C etc.  
 C  $y(5) = E$  for component 1  
 C  $y(6) = A$  for component 2  
 C etc.

C  $fd(v)$  is a vector of feedstream flows  
 C  $s(v)$  is a vector of sidestream flows

C the liquid and vapor enthalpies are fit to straight lines:  
 C  $HL = YV*T + BL$   
 C  $HV = YV*T + BV$   
 C  $YV(cp)$ ,  $YL(cp)$ ,  $BV(cp)$ ,  $BL(cp)$  hold these parameters  
 C data (2) Holds user options data  
 C sum is the sum of the squares of f  
 C xnew(v) is the vector of delta x

- C ctemp(cvar,cvar) is the reduced Jacobian
- C ftemp(cvar) is the reduced discrepancy vector
- C xtemp(cvar) is the reduced delta x vector
- C dist(7) contains user option information

parameter (v = 468, trav = 52, cp = 4, cvar = 30)

iter = 0

- C open input and output files

```
open(unit = 70, file = 'data.dat', status = 'old')
open(unit = 60, file = 'dist.dat', status = 'old')
open(unit = 20, file = 'input.dat', status = 'old')
open(unit = 45, file = 's.dat', status = 'old')
open(unit = 55, file = 'fd.dat', status = 'old')
open(unit = 30, file = 'dist1.dat', status = 'old')
open(unit = 95, file = 'x.out', status = 'new')
```

- C read in all the input data

```
call input(n,y,yv,yl,bv,bl,dist,data,mxiter,s,fd,x,choice)
```

- C compiler directive to get all four CPUs

CMICS GETCPUS 4

- C Set up initial discrepancy vector

```
call discrep(f,x,s,fd,n,y,bv,bl,yv,yl,dist,data)
```

- C Set up initial derivative matrix

```
10 call deriv(f,x,s,fd,n,y,c,bv,bl,yv,yl,dist,data)
```

- C Decide if user wants to solve using regular block Gaussian
- C elimination (REGSOLVE) or parallel elimination (PARTIAL.)

```
if (choice .eq. 1)then
  call regsolve(c,f,xnew)
  goto 200
endif
```

- C If parallel mode is selected, then do a partial, parallel
- C Gaussian elimination

call partial(c,f)

- C select the lines of the system that border the processor
- C divisions

do 12 part = 1,3  
call select(f,c,ctemp,ftemp,part)

12 continue

- C solve the reduced matrix

call fstsolve(ctemp,ftemp,xtemp)

- C perform back substitution

do 13 part = 1,3  
call scatter1(xtemp,xnew,part)

13 continue

call scatter2(f,c,xtemp,xnew)

- C find the best fractional part of the delta x's

200 call search(x,xnew,fd,s,f,n,y,bv,bl,yv,yl,dist,data)

- C make sure no variables are in a physically meaningless
- C region

call realvar(x)

- C determine the sum of the squares of the discrepancy
- C functions

call discrep(f,x,s,fd,n,y,bv,bl,yv,yl,dist,data)

call sum (f,sums)

- C determine if the method has converged

iter = iter + 1  
if (iter .eq. mxiter .or. sums .lt. tol) then  
goto 20  
else  
goto 10  
endif

C print out final results

20 call out(x,sums,iter)

stop

end

C

C subroutine DISCREP evaluates the discrepancy

C functions using the current values for the variables

C

subroutine DISCREP(f,x,s,fd,n,y,bv,bl,yv,yl,dist,data)

integer i,j,m,tray,cp,dist(7),opt1,opt2,v

real f(468),x(-4:473),n(52),y(20),bv(4),bl(4),data(2),

+ s(468),fd(468),yv(4),yl(4),ev,el,evp,ell,qv,ql,qvp,vptotc,

+ vtotc,vtotr,ltotc,ltotr,sum,cev,cel,cevp,rev,rel,rell,fsum

parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C dimensioning information

C

C f(v)

C x(-cp:v+cp+1)

C n(tray)

C y(5\*cp)

C bv(cp),bl(4),yv(4),yl(4)

C data(2)

C s(v),fd(v)

opt1 = 0

opt2 = 0

C Determine if the problem is and absorber or distillation

C column. If absorber, then goto 5

if (dist(1) .eq. 0) then

goto 5

endif

C This section does the condenser and reboiler

C opt1 and opt2 are set to one when there are reboilers and

C condensers. That way, the main loop knows not to evaluate

C the first and last trays in the normal fashion.

```
opt1 = 1
opt2 = 1
```

- C Evaluate the energy line for the condenser and reboiler  
 C based on what user options have been selected

```
if (dist(4) .eq. 1) then
  if (dist(2) .eq. 1) then

    cev = 0.0
    cel = 0.0
    cevp = 0.0
    fsum = 0.0
    do 10 i = 1,cp
      cev = cev + (x(i)+s(i))*(yl(i)*x(cp+1)+bl(i))
      cel = cel + (x(i+cp+1)+s(i+cp+1))*(yl(i)*x(cp+1)+bl(i))
      cevp = cevp + x(i+2*cp+1)*(yv(i)*x(3*cp+2)+bv(i))
      fsum = fsum + fd(i)*(yv(i)*fd(cp+1)+bv(i))
      +      + fd(i+cp+1)*(yl(i)*fd(cp+1)+bl(i))
10    continue

    f(cp+1) = -(cev + cel - cevp - fsum + data(1))

  else

    cev = 0.0
    cel = 0.0
    cevp = 0.0
    fsum = 0.0
    do 20 i = 1,cp
      cev = cev + (x(i)+s(i))*(yv(i)*x(cp+1)+bv(i))
      cel = cel + (x(i+cp+1)+s(i+cp+1))*(yl(i)*x(cp+1)+bl(i))
      cevp = cevp + x(i+2*cp+1)*(yv(i)*x(3*cp+2)+bv(i))
      fsum = fsum + fd(i)*(yv(i)*fd(cp+1)+bv(i))
      +      + fd(i+cp+1)*(yl(i)*fd(cp+1)+bl(i))
20    continue

    f(cp+1) = -(cev + cel -cevp - fsum + data(1))

  endif
endif

if (dist(5) .eq. 1) then
  If (dist(3) .eq. 1) then
```



```

    rev = 0.0
    rel = 0.0
    rell = 0.0
    fsum = 0.0
    do 30 i = 1,cp
        rev = rev + (x(v-2*cp-1+i)+s(v-2*cp-1+i))
+         *(yv(i)*x(v-cp)+bv(i))
        rel = rel + (x(v-cp+i)+s(v-cp+i)) *(yv(i)*x(v-cp)+bv(i))
        rell = rell + x(v-3*cp-1+i)*(yl(i)*x(v-3*cp-1)+bl(i))
        fsum = fsum + fd(v-2*cp-1+i)*(yv(i)*fd(v-cp)+bv(i))
+         + fd(v-cp+i)*(yl(i)*fd(v-cp)+bl(i))
30    continue

    f(v-cp) = -(rev + rel - rell - fsum - data(2))

else

    rev = 0.0
    rel = 0.0
    rell = 0.0
    fsum = 0.0
    do 40 i = 1,cp
        rev = rev + (x(v-2*cp-1+i)+s(v-2*cp-1+i))
+         *(yv(i)*x(v-cp)+bv(i))
        rel = rel + (x(v-cp+i)+s(v-cp+i))*(yl(i)*x(v-cp)+bl(i))
        rell = rell + x(v-3*cp-1+i)*(yl(i)*x(v-3*cp-1)+bl(i))
        fsum = fsum + fd(v-2*cp-1+i)*(yv(i)*fd(v-cp)+bv(i))
+         + fd(v-cp+i)*(yl(i)*fd(v-cp)+bl(i))
40    continue

    f(v-cp) = -(rev + rel - rell - fsum - data(2))

endif

endif

vtotc = 0.0
ltotc = 0.0
vtotr = 0.0
ltotr = 0.0
vptotc = 0.0
do 50 i = 1,cp
    vtotc = vtotc + x(i)
    ltotc = ltotc + x(cp+1+i)
    vtotr = vtotr + x(v-2*cp-1+i)

```

```

    ltotr = ltotr + x(v-cp+i)
    vptotc = vptotc + x(2*cp+1+i)
50  continue

    if(dist(4) .eq. 2) then
        f(cp+1) = -(ltotc - vtotc*data(1))
    endif

    if (dist(4) .eq. 3) then
        f(cp+1) = -(vtotc - data(1))
    endif

    If (dist(4) .eq. 4) then
        f(cp+1) = -(x(dist(6)) - data(1))
    endif

    if (dist(4) .eq. 5) then
        f(cp+1) = -(x(cp+1) - data(1))
    endif

    if(dist(5) .eq. 2) then
        f(v-cp) = -(vtotr - ltotr*data(2))
    endif

    if (dist(5) .eq. 3) then
        f(v-cp) = -(ltotr - data(2))
    endif

    If (dist(5) .eq. 4) then
        f(v-cp) = -(x(v-cp+dist(7)) - data(2))
    endif

    if (dist(5) .eq. 5) then
        f(v-cp) = -(x(v-cp) - data(2))
    endif

C    evaluate the mass and equilibrium lines for the
C    condenser and reboiler based upon user options

    if(dist(2) .eq. 1) then

        sum = 0.0
        do 60 i = 1,cp
            sum = sum + (y(5*(i-1)+1)*x(cp+1))**4

```

```

+      + y(5*(i-1)+2)*x(cp+1)**3
+      + y(5*(i-1)+3)*x(cp+1)**2
+      + y(5*(i-1)+4)*x(cp+1)
+      + y(5*(i-1)+5))
+      *x(cp+1+i)/ltotc
60  continue

      do 70 i = 1,cp
        f(i) = -(x(i) + x(i+cp+1) - x(i-cp) - x(i+2*cp+1)
+          - fd(i) - fd(i+cp+1) + s(i) + s(i+cp+1))
        f(i+cp+1) = -(x(i+cp+1) - ltotc*x(i)/vtotc)
70  continue

      f(cp+2) = -(1 - sum)

    else

      do 80 i = 1,cp
        f(i) = -(x(i) + x(i+cp+1) - x(i-cp) - x(i+2*cp+1)
+          - fd(i) - fd(i+cp+1) + s(i) + s(i+cp+1))
        f(i+cp+1) = -(n(1)*(y(5*(i-1)+1)*x(cp+1)**4
+          + y(5*(i-1)+2)*x(cp+1)**3
+          + y(5*(i-1)+3)*x(cp+1)**2
+          + y(5*(i-1)+4)*x(cp+1)
+          + y(5*(i-1)+5))
+          *vtotc*x(i+cp+1)/ltotc
+          - x(i) + (1-n(1))*x(i+2*cp+1)*vtotc/vptotc)
80  continue

    endif

    if(dist(3) .eq. 1) then

      sum = 0.0
      do 90 i = 1,cp
        sum = sum + (y(5*(i-1)+1)*x(v-cp)**4
+          + y(5*(i-1)+2)*x(v-cp)**3
+          + y(5*(i-1)+3)*x(v-cp)**2
+          + y(5*(i-1)+4)*x(v-cp)
+          + y(5*(i-1)+5))
+          *vtotr/x(v-2*cp-1+i)
90  continue

```

```

do 100 i = 1,cp
  f(v-2*cp-1+i)=- (x(v-2*cp-1+i)+x(v-cp+i)-x(v-3*cp-1+i)-x(v+i)
+   -fd(v-2*cp-1+i)-fd(v-cp+i)+s(v-2*cp-1+i)+s(v-cp+i))
  f(v-cp+i)=- (x(v-2*cp-1+i) - vtotr*x(v-cp+i)/ltotr)
100 continue

  f(v-cp+1) = -(1 - sum)

else

  do 110 i = 1,cp
    f(v-2*cp-1+i)=- (x(v-2*cp-1+i)+x(v-cp+i)-x(v-3*cp-1+i)-x(v+i)
+   -fd(v-2*cp-1+i)-fd(v-cp+i)+s(v-2*cp-1+i)+s(v-cp+i))
    f(v-cp+i)=- (n(tray)*(y(5*(i-1)+1)*x(v-cp)**4
+   + y(5*(i-1)+2)*x(v-cp)**3
+   + y(5*(i-1)+3)*x(v-cp)**2
+   + y(5*(i-1)+4)*x(v-cp)
+   + y(5*(i-1)+5))
+   *vtotr*x(v-cp+i)/ltotr
+   - x(v-2*cp-1+i))
110 continue

  endif

  if (dist(4) .eq. 2 .or. dist(4) .eq. 3 .or. dist(4) .eq. 4)then
    dump = f(cp+1)
    f(cp+1) = f(cp+2)
    f(cp+2) = dump
  endif

  if (dist(5) .eq. 2 .or. dist(5) .eq. 3 .or. dist(5) .eq. 4)then
    dump = f(v-cp)
    f(v-cp) = f(v-cp+1)
    f(v-cp+1) = dump
  endif

```

C this is the main loop that evaluates the trays. If the problem  
 C is an absorber, then opt1 and opt2 will be zero and it will  
 C iterate from 1 to tray. Otherwise, it will iterate from  
 C 2 to tray-1.

```

5 do 120 j = 1+opt1,tray-opt2

```

```

ev = 0.0
el = 0.0
evp = 0.0
ell = 0.0
fsum = 0.0
qv = 0.0
qvp = 0.0
ql = 0.0
m = (j-1)*(2*cp+1)
do 130 i = 1,cp

    ev = ev + (x(i+m)+s(i+m))*(yv(i)*x(m+cp+1) + bv(i))
    el = el + (x(i+m+cp+1)+s(i+m+cp+1))*(yl(i)*x(m+cp+1)+bl(i))
    evp = evp + x(i+m+2*cp+1)*(yv(i)*x(m+3*cp+2) + bv(i))
    ell = ell + x(i+m-cp)*(yl(i)*x(m-cp) + bl(i))
    fsum = fsum + fd(i+m)*(yv(i)*fd(m+cp+1)+bv(i))
+       + fd(i+m+cp+1)*(yl(i)*fd(m+cp+1)+bl(i))

    qv = qv + x(i+m)
    ql = ql + x(i+m+cp+1)
    qvp = qvp + x(i+m+2*cp+1)

130  continue

f(m+cp+1) = -(el + ev - fsum - ell - evp)

do 140 i = 1,cp

    f(i+m) = -(x(i+m)+x(i+m+cp+1)-x(i+m-cp)-x(i+m+2*cp+1)
+       -fd(i+m)-fd(i+m+cp+1)+s(i+m)+s(i+m+cp+1))
    f(i+m+cp+1) = -(n(j)*(y(5*(i-1)+1)*x(m+cp+1)**4
+       + y(5*(i-1)+2)*x(m+cp+1)**3
+       + y(5*(i-1)+3)*x(m+cp+1)**2
+       + y(5*(i-1)+4)*x(m+cp+1)
+       + y(5*(i-1)+5))
+       *qv*x(i+m+cp+1)/ql
+       - x(i+m) + (1-n(j))*x(i+m+2*cp+1)*qv/qvp)

140  continue

120  continue

return
end

```

C subroutine DERIV evaluates the Jacobian at the present  
C values for the variables.

C the compiler directive CMICS MICRO designates this  
C subroutine as one in which microtasking will occur

# CMICS MICRO

subroutine DERIV(f,x,s,fd,n,y,c,bv,bl,yv,yl,dist,data)

integer i,j,k,l,m,o,tray,cp,v,dist(7),opt1,opt2  
real f(468), x(-4:473), c(468,-9:477), n(52), y(20),  
+ bv(4), bl(4), yv(4), yl(4), sum, bbl, bbv, qbl,  
+ qbv, qbvp, option, data(2), s(468), fd(468),dump(18)  
parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C dimensioning information

C

C f(v),s(v),fd(v)

C x(-cp:v+cp+1)

C c(v,-2\*cp-1:v+2\*cp+1)

C n(tray)

C y(5\*cp)

C bv(4),bl(4),yv(4),yl(4)

C data(2)

C dump(4\*cp+2)

C again determine if the problem is an absorber or distillation

C problem

opt1 = 0

opt2 = 0

if (dist(1) .eq. 0) then

goto 5

endif

C Condenser

opt1 = 1

opt2 = 1

C B-matrix

if (dist(4) .eq. 1) then

```

if (dist(2) .eq. 1) then

  do 10 j = 1,cp
    c(cp+1,j) = yl(j)*x(cp+1) + bl(j)
    c(cp+1,j+cp+1) = yl(j)*x(cp+1) + bl(j)
10  continue

  bbl = 0.0
  bbv = 0.0

  do 20 i = 1,cp
    bbl = bbl + (x(i+cp+1)+s(i+cp+1))*yl(i)
    bbv = bbv + (x(i)+s(i))*yl(i)
20  continue

  c(cp+1,cp+1) = bbl+bbv

else

  do 30 j = 1,cp
    c(cp+1,j) = yv(j)*x(cp+1) + bv(j)
    c(cp+1,cp+1+j) = yl(j)*x(cp+1) + bl(j)
30  continue

  bbl = 0.0
  bbv = 0.0
  do 40 i = 1,cp
    bbl = bbl + (x(i+cp+1)+s(i+cp+1))*yl(i)
    bbv = bbv + (x(i)+s(i+cp+1))*yv(i)
40  continue
  c(cp+1,cp+1) = bbl+bbv

endif

elseif (dist(4) .eq. 2) then

  do 50 j = 1,cp
    c(cp+1,j) = -data(1)
50  continue

  c(cp+1,cp+1) = 0.0

  do 60 j = cp+2,2*cp+1
    c(cp+1,j) = 1.0
60  continue

```

```

elseif(dist(4) .eq. 3) then

    do 70 j = 1,cp
        c(cp+1,j) = 1.0
70    continue

    do 80 j = cp+1,2*cp+1
        c(cp+1,j) = 0.0
80    continue

elseif (dist(4) .eq. 4) then

    do 90 j = 1,cp
        if ( j .eq. dist(6)) then
            c(cp+1,j) = 1.0
        else
            c(cp+1,j) = 0.0
        endif
90    continue

    do 100 j = cp+1,2*cp+1
        c(cp+1,j) = 0.0
100    continue

elseif (dist(4) .eq.5) then

    do 110 j = 1,2*cp+1
        c(cp+1,j) = 0.0
110    continue

    c(cp+1,cp+1) = 1.0

endif

do 120 j = 1,cp
    do 130 i = 1,cp
        if(i .eq. j) then
            c(i,j) = 1.0
        else
            c(i,j) = 0.0
        endif
130    continue
120    continue

```



```

do 140 i = 1,cp
  c(i,cp+1) = 0.0
140 continue

do 150 j = cp+2,2*cp+1
  do 160 i = 1,cp
    if((j-cp-1).eq. i) then
      c(i,j) = 1.0
    else
      c(i,j) = 0.0
    endif
  160 continue
150 continue

  if (dist(2).eq. 1) then

    qbl = 0.0
    qbvp = 0.0
    qbv = 0.0

    do 170 i = 1,cp
      qbl = qbl + x(i+cp+1)
      qbvp = qbvp + x(i+2*cp+1)
      qbv = qbv + x(i)
    170 continue

    do 180 j = 1,cp
      c(cp+2,j) = 0.0
    180 continue

    do 190 j = 1,cp
      do 200 i = cp+3,2*cp+1
        if ((i-cp-1).eq. j) then
          c(i,j) = -qbl*(qbv-x(i-cp-1))/(qbv*qbv)
        else
          c(i,j) = -qbl*x(i-cp-1)/(qbv*qbv)
        endif
      200 continue
    190 continue

    sum = 0.0
    do 210 i = 1,4
      sum = sum - (4*y(5*(i-1)+1)*x(cp+1)**3
+      + 3*y(5*(i-1)+2)*x(cp+1)**2
+      + 2*y(5*(i-1)+3)*x(cp+1)

```

```

+      + y(5*(i-1)+4))
+      *x(i+cp+1)/qbl
210  continue

      c(cp+2,cp+1) = sum

do 215 i = 2,cp
      c(cp+1+i,cp+1) = 0.0
215  continue

      sum = 0.0
do 220 j = cp+2,2*cp+1
      sum = sum + (y(5*(j-cp-2)+1)*x(cp+1)**4
+      + y(5*(j-cp-2)+2)*x(cp+1)**3
+      + y(5*(j-cp-2)+3)*x(cp+1)**2
+      + y(5*(j-cp-2)+4)*x(cp+1)
+      + y(5*(j-cp-2)+5))
+      *x(j)/(qbl*qbl)
220  continue

do 230 j = cp+2,2*cp+1
      c(cp+2,j) = (y(5*(j-cp-2)+1)*x(cp+1)**4
+      + y(5*(j-cp-2)+2)*x(cp+1)**3
+      + y(5*(j-cp-2)+3)*x(cp+1)**2
+      + y(5*(j-cp-2)+4)*x(cp+1)
+      + y(5*(j-cp-2)+5))
+      *(x(j)-qbl)
+      /(qbl*qbl) + sum -
+      (y(5*(j-cp-2)+1)*x(cp+1)**4
+      + y(5*(j-cp-2)+2)*x(cp+1)**3
+      + y(5*(j-cp-2)+3)*x(cp+1)**2
+      + y(5*(j-cp-2)+4)*x(cp+1)
+      + y(5*(j-cp-2)+5))
+      *x(j)/(qbl*qbl)
230  continue

do 240 j = cp+2,2*cp+1
do 250 i = cp+3,2*cp+1
      if (i .eq. j) then
        c(i,j) = 1-x(i-cp-1)/qbv
      else
        c(i,j) = -x(i-cp-1)/qbv
      endif
250  continue
240  continue

```

```

else
  qbl = 0.0
  qbvp = 0.0
  qbv = 0.0

  do 260 i = 1,cp
    qbl = qbl + x(i+cp+1)
    qbvp = qbvp + x(i+2*cp+1)
    qbv = qbv + x(i)
260  continue

  do 270 j = 1,cp
    do 280 i = cp+2,2*cp+1

      if((i-cp-1) .eq. j)then
        option = 1.0
      else
        option = 0.0
      endif

      c(i,j) = n(1)*(y(5*(i-cp-2)+1)*x(cp+1)**4
+             + y(5*(i-cp-2)+2)*x(cp+1)**3
+             + y(5*(i-cp-2)+3)*x(cp+1)**2
+             + y(5*(i-cp-2)+4)*x(cp+1)
+             + y(5*(i-cp-2)+5))
+             *x(i)/qbl - option + (1-n(1))
+             *x(i+cp)/qbvp

280  continue
270  continue

  do 290 i = 1,cp
    c(i+cp+1,cp+1) = n(1)*(4*y(5*(i-1)+1)*x(cp+1)**3
+             +3*y(5*(i-1)+2)*x(cp+1)**2
+             +2*y(5*(i-1)+3)*x(cp+1)
+             + y(5*(i-1)+4))
+             *qbv*x(i+cp+1)/qbl
290  continue

  do 300 j = cp+2,2*cp+1
    do 310 i = cp+2,2*cp+1

      if(i .eq. j)then
        option = qbl*n(1)*(y(5*(i-cp-2)+1)*x(cp+1)**4
+             + y(5*(i-cp-2)+2)*x(cp+1)**3

```

```

+          + y(5*(i-cp-2)+3)*x(cp+1)**2
+          + y(5*(i-cp-2)+4)*x(cp+1)
+          + y(5*(i-cp-2)+5))
+          *qbv

    else
      option = 0.0
    endif

    c(i,j) = (option-n(1)*(y(5*(i-cp-2)+1)*x(cp+1)**4
+          + y(5*(i-cp-2)+2)*x(cp+1)**3
+          + y(5*(i-cp-2)+3)*x(cp+1)**2
+          + y(5*(i-cp-2)+4)*x(cp+1)
+          + y(5*(i-cp-2)+5))
+          *qbv*x(i))/(qbl*qbl)

310    continue
300    continue

endif

```

## C C matrix

```

if (dist(4) .eq. 1) then

  do 320 j = 1,cp
    c(cp+1,2*cp+1+j) = -(yv(j)*x(3*cp+2)+bv(j))
320  continue

    sum = 0.0
  do 330 i = 1,cp
    sum = sum - x(2*cp+1+i)*yv(i)
330  continue

    c(cp+1,3*cp+2) = sum

  else

    do 340 j = 1,cp+1
      c(cp+1,2*cp+1+j) = 0.0
340  continue

  endif

```

```

do 350 j = 1,cp
  do 360 i = 1,cp
    if(i .eq. j)then
      c(i,2*cp+1+j) = -1.0
    else
      c(i,2*cp+1+j) = 0.0
    endif
360   continue
350   continue

  if (dist(2) .eq. 0) then

    do 370 i = cp+2,2*cp+1
      do 380 j = 1,cp

        if((i-cp-1) .eq. j)then
          option = qbvp*(1-n(1))*qbv
        else
          option = 0.0
        endif

        c(i,2*cp+1+j) = (option - (1-n(1))*x(i+cp)
+          *qbv)/(qbvp*qbvp)
380   continue
370   continue

      else

        do 390 i = 1,cp
          do 400 j = 1,cp
            c(i,2*cp+1+j) = 0.0
            c(cp+1+i,2*cp+1+j) = 0.0
400   continue
390   continue

          endif

          do 410 i = 1,cp
            c(i,3*cp+2) = 0.0
            c(cp+1+i,3*cp+2) = 0.0
410   continue

          do 420 j = cp+2,2*cp+1
            do 430 i = 1,2*cp+1

```

```

      c(i,2*cp+1+j) = 0.0
430  continue
420  continue

```

C reboiler

C A-matrix

```

      do 450 j = 1,cp
        do 460 i = 1,2*cp+1
          c(v-2*cp-1+i,j+v-4*cp-2) = 0.0
460  continue
450  continue

```

if (dist(5) .eq. 1)then

```

      sum = 0.0
      do 470 j = 1 cp
        c(v-cp,v-3*cp-1+j) = -(yl(j)*x(v-3*cp-1)+bl(j))
        sum = sum - x(v-j-3*cp-1)*yl(j)
470  continue

```

c(v-cp,v-3\*cp-1) = sum

else

```

      do 471 j = 1,2*cp+1
        c(v-cp,v-4*cp-2+j) = 0.0
471  continue

```

endif

```

      do 480 i = 2,2*cp+1
        c(v-cp-1+i,v-3*cp-1) = 0.0
480  continue

```

```

      do 490 j = cp+2,2*cp+1
        do 500 i = cp+2,2*cp+1
          c(v-cp-1+i,j+v-4*cp-2) = 0.0
500  continue
490  continue

```

```

      do 510 j = cp+2,2*cp+1
        do 520 i = 1,cp

```

```

        if((j-cp-1).eq. i)then
            c(v-2*cp-1+i,j+v-4*cp-2) = -1.0
        else
            c(v-2*cp-1+i,j+v-4*cp-2) = 0.0
        endif
520    continue
510    continue

```

### C B-matrix

```

if (dist(5).eq. 1) then

    if (dist(3).eq. 1) then

        do 530 j = 1,cp
            c(v-cp,v-2*cp-1+j) = yv(j)*x(v-cp) + bv(j)
            c(v-cp,v-cp+j) = yv(j)*x(v-cp) + bv(j)
530        continue

        bbl = 0.0
        bbv = 0.0

        do 540 i = 1,cp
            bbl = bbl + (x(v-cp+i)+s(v-cp+i))*yv(i)
            bbv = bbv + (x(v-2*cp-1+i)+s(v-2*cp-1+i))*yv(i)
540        continue

        c(v-cp,v-cp) = bbl+bbv

    else

        do 550 j = 1,cp
            c(v-cp,v-2*cp-1+j) = yv(j)*x(v-cp) + bv(j)
            c(v-cp,v-cp+j) = yl(j)*x(v-cp) + bl(j)
550        continue

        bbl = 0.0
        bbv = 0.0

        do 560 i = 1,cp
            bbl = bbl + (x(v-cp+i)+s(v-cp+i))*yl(i)
            bbv = bbv + (x(v-2*cp-1+i)+s(v-cp+i))*yv(i)
560        continue

```

```

        c(v-cp,v-cp) = bbl+bbv
endif

elseif (dist(5) .eq. 2) then

    do 570 j = 1,cp
        c(v-cp,v-2*cp-1+j) = 1.0
570    continue

        c(v-cp,cp+1) = 0.0

        do 580 j = cp+2,2*cp+1
            c(v-cp,v-2*cp-1+j) = data(2)
580    continue

elseif(dist(5) .eq. 3) then

    do 590 j = 1,cp+1
        c(v-cp,v-2*cp-1+j) = 0.0
590    continue

        do 600 j = cp+2,2*cp+1
            c(v-cp,v-2*cp-1+j) = 1.0
600    continue

elseif (dist(5) .eq. 4) then

    do 610 j = 1,cp
    if ( j .eq. dist(7)) then
        c(v-cp,v-cp+j) = 1.0
    else
        c(v-cp,v-cp+j) = 0.0
    endif
610    continue

        do 620 j = 1,5
            c(v-cp,v-2*cp-1+j) = 0.0
620    continue

elseif (dist(5) .eq.5) then

    do 630 j = 1,2*cp+1
        c(v-cp,v-2*cp-1+j) = 0.0
630    continue

```



```

      c(v-cp,v-cp) = 1.0

endif

do 640 j = 1,cp
  do 650 i = 1,cp
    if(i .eq. j) then
      c(v-2*cp-1+i,v-2*cp-1+j) = 1.0
    else
      c(v-2*cp-1+i,v-2*cp-1+j) = 0.0
    endif
650   continue
640   continue

    do 660 i = 1,cp
      c(v-2*cp-1+i,v-cp) = 0.0
660   continue

    do 670 j = cp+2,2*cp+1
      do 680 i = 1,cp
        if((j-cp-1) .eq. i) then
          c(v-2*cp-1+i,v-2*cp-1+j) = 1.0
        else
          c(v-2*cp-1+i,v-2*cp-1+j) = 0.0
        endif
680   continue
670   continue

    if (dist(3) .eq. 1) then

      qbl = 0.0
      qbv = 0.0

      do 690 i = 1,cp
        qbl = qbl + x(v-cp+i)
        qbv = qbv + x(v-2*cp-1+i)
690   continue

      do 700 j = cp+2,2*cp+1
        c(v-cp+1,v-2*cp-1+j) = 0.0
700   continue

      do 710 j = cp+2,2*cp+1
        do 720 i = cp+3,2*cp+1
          if ((i-1) .eq. j) then

```

```

      c(v-2*cp-1+i,v-2*cp-1+j)=-qbv*(qbl-x(v-2*cp-1+i))
+      /(qbl*qbl)
    else
      c(v-2*cp-1+i,v-2*cp-1+j)=-qbv*x(v-2*cp-1+i)/(qbl*qbl)
    endif
720  continue
710  continue

    sum = 0.0
    do 730 i = 1,4
      sum = sum - (4*y(5*(i-1)+1)*x(v-cp)**3
+      + 3*y(5*(i-1)+2)*x(v-cp)**2
+      + 2*y(5*(i-1)+3)*x(v-cp)
+      + y(5*(i-1)+4))
+      *qbv/x(v-2*cp-1+i)
730  continue

    c(v-cp+1,v-cp) = sum

    do 740 i = 2,cp
      c(v-cp+i,v-cp) = 0.0
740  continue

    sum = 0.0
    do 750 j = 1,cp
      sum = sum + (y(5*(j-1)+1)*x(v-cp)**4
+      + y(5*(j-1)+2)*x(v-cp)**3
+      + y(5*(j-1)+3)*x(v-cp)**2
+      + y(5*(j-1)+4)*x(v-cp)
+      + y(5*(j-1)+5))
+      /x(v-2*cp-1+j)
750  continue

    do 760 j = 1,cp
      c(v-cp+1,v-2*cp-1+j) = (y(5*(j-1)+1)*x(v-cp)**4
+      + y(5*(j-1)+2)*x(v-cp)**3
+      + y(5*(j-1)+3)*x(v-cp)**2
+      + y(5*(j-1)+4)*x(v-cp)
+      + y(5*(j-1)+5))
+      *(qbv-x(v-2*cp-1+j))
+      /(qbv*qbv) + sum -
+      (y(5*(j-1)+1)*x(v-cp)**4
+      + y(5*(j-1)+2)*x(v-cp)**3
+      + y(5*(j-1)+3)*x(v-cp)**2
+      + y(5*(j-1)+4)*x(v-cp)

```

```

+          + y(5*(j-1)+5))
+          /x(v-2*cp-1+j)
760  continue

      do 770 j = 1,cp
        do 780 i = cp+3,2*cp+1
          if (i-cp-2 .eq. j) then
            c(v-2*cp-1+i,v-2*cp-1+j) = 1-x(v-2*cp-1+i)/qbl
          else
            c(v-2*cp-1+i,v-2*cp-1+j) = -x(v-2*cp-1+i)/qbl
          endif
670  continue
770  continue

      else

        qbl = 0.0
        qbvp = 0.0
        qbv = 0.0

        do 790 i = 1,cp
          qbl = qbl + x(v-cp+i)
          qbvp = qbvp + x(v+i)
          qbv = qbv + x(v-2*cp-1+i)
790  continue

        do 800 j = 1,cp
          do 810 i = cp+2,2*cp+1

            if((i-cp-1) .eq. j)then
              option = 1.0
            else
              option = 0.0
            endif

            c(v-2*cp-1+i,v-2*cp-1+j)=n(1)*(y(5*(i-cp-2)+1)*x(v-cp)**4
+              + y(5*(i-cp-2)+2)*x(v-cp)**3
+              + y(5*(i-cp-2)+3)*x(v-cp)**2
+              + y(5*(i-cp-2)+4)*x(v-cp)
+              + y(5*(i-cp-2)+5))
+              *x(v-2*cp-1+i)/qbl - option

810  continue
800  continue

```

```

do 820 i = 1,cp
  c(v-cp+i,v-cp) = n(1)*(4*y(5*(i-1)+1)*x(v-cp)**3
+                + 3*y(5*(i-1)+2)*x(v-cp)**2
+                + 2*y(5*(i-1)+3)*x(v-cp)
+                + y(5*(i-1)+4))
+                *qbv*x(v-cp+i)/qbl
820  continue

do 830 j = cp+2,2*cp+1
  do 840 i = cp+2,2*cp+1

    if(i .eq. j)then
      option = qbl*n(1)*(y(5*(i-cp-2)+1)*x(v-cp)**4
+                + y(5*(i-cp-2)+2)*x(v-cp)**3
+                + y(5*(i-cp-2)+3)*x(v-cp)**2
+                + y(5*(i-cp-2)+4)*x(v-cp)
+                + y(5*(i-cp-2)+5))
+                *qbv
    else
      option = 0.0
    endif

    c(v-2*cp-1+i,v-2*cp-1+j)=(option-n(1)*
+                (y(5*(i-cp-2)+1)*x(v-cp)**4
+                + y(5*(i-cp-2)+2)*x(v-cp)**3
+                + y(5*(i-cp-2)+3)*x(v-cp)**2
+                + y(5*(i-cp-2)+4)*x(v-cp)
+                + y(5*(i-cp-2)+5))
+                *qbv*x(v-2*cp-1+i))/(qbl*qbl)

840  continue
830  continue

  endif

  if(dist(4) .eq. 2 .or. dist(4) .eq. 3 .or. dist(4) .eq. 4)then
    do 831 i = 1,4*cp+2
      dump(i) = c(cp+1,i)
831  continue
      do 832 i = 1,4*cp+2
        c(cp+1,i) = c(cp+2,i)
832  continue
      do 833 i = 1,4*cp+2
        c(cp+2,i) = dump(i)

```

```

833  continue
      endif

      if(dist(5) .eq. 2 .or. dist(5) .eq. 3 .or. dist(5) .eq. 4)then
        do 834 i = 1,4*cp+2
          dump(i) = c(v-cp,v-4*cp-2+i)
834    continue
          do 835 i = 1,4*cp+2
            c(v-cp,v-4*cp-2+i) = c(v-cp+1,v-4*cp-2+i)
835    continue
          do 836 i = 1,4*cp+2
            c(v-cp+1,v-4*cp-2+i) = dump(i)
836    continue
          endif

```

```

5    sum = 0.0

```

C this is the main loop of the program. a few explanations are  
 C in order here. First, note CMIC\$ DO GLOBAL. This tells the  
 C compiler that every iteration of loop 85 is independent of  
 C each other. therefore, it should divide the iterations of the  
 C loop among the four processors. Also, note that the counter  
 C from 1,52 rather than 1+Opt1,1-opt2. This is done because  
 C the compiler can't microtask the loop if the number of iterations  
 C is a variable. It has to be some constant. (This particular  
 C problem had 52 trays)

```

CMIC$ DO GLOBAL

```

```

C  loop 850 from 1 to tray

```

```

      do 850 k = 1,52

```

```

C  A-matrix

```

```

      m = (k-1)*(2*cp+1)

```

```

      do 870 j = 1,cp
        do 880 i = 1,2*cp+1
          c(m+i,m+j-2*cp-1) = 0.0

```

```

880    continue

```

```

870    continue

```

```

      sum = 0.0

```

```

      do 890 j = 1,cp

```

```

      c(m+cp+1,m+j-cp) = -(yl(j)*x(m-cp)+bl(j))
      sum = sum - x(m+j-cp)*yl(j)
890  continue

      c(m+cp+1,m-cp) = sum

      do 900 i = 1,cp
        c(m+i,m-cp) = 0.0
900  continue

      do 910 j = cp+1,2*cp+1
        do 920 i = cp+2,2*cp+1
          c(m+i,m+j-2*cp-1) = 0.0
920  continue
910  continue

      do 930 j = cp+2,2*cp+1
        do 940 i = 1,cp
          if((j-cp-1).eq. i)then
            c(m+i,m+j-2*cp-1) = -1.0
          else
            c(m+i,m+j-2*cp-1) = 0.0
          endif
940  continue
930  continue

860  sum = 0.0

850  continue

CC  B-matrix

CMICS DO GLOBAL

C  loop 950 from 1 to tray

      do 950 k = 1,52

        m = (k-1)*(2*cp+1)

        do 960 j = 1,cp
          c(m+cp+1,m+j) = yv(j)*x(m+cp+1) + bv(j)
          c(m+cp+1,m+j+cp+1) = yl(j)*x(m+cp+1) + bl(j)
960  continue

```

```

    bbl = 0.0
    bbv = 0.0

    do 970 i = 1,cp
        bbl = bbl + (x(m+i+cp+1)+s(m+i+cp+1))*yl(i)
        bbv = bbv + (x(m+i)+s(m+i))*yv(i)
970    continue

    c(m+cp+1,m+cp+1) = bbl+bbv

    do 980 j = 1,cp
        do 990 i = 1,cp
            if(i .eq. j)then
                c(m+i,m+j) = 1.0
            else
                c(m+i,m+j) = 0.0
            endif
990    continue
980    continue

    do 1000 i = 1,cp
        c(m+i,m+cp+1) = 0.0
1000    continue

    do 1010 j = cp+2,2*cp+1
        do 1020 i = 1,cp
            if((j-cp-1) .eq. i)then
                c(m+i,m+j) = 1.0
            else
                c(m+i,m+j) = 0.0
            endif
1020    continue
1010    continue

    qbl = 0.0
    qbvp = 0.0
    qbv = 0.0

    do 1030 i = 1,cp
        qbl = qbl + x(m+i+cp+1)
        qbvp = qbvp + x(m+i+2*cp+1)
        qbv = qbv + x(m+i)
1030    continue

```

```

do 1040 j = 1,cp
  do 1050 i = cp+2,2*cp+1

    if((i-cp-1) .eq. j)then
      option = 1.0
    else
      option = 0.0
    endif

    c(m+i,m+j) = n(k)*(y(5*(i-cp-2)+1)*x(m+cp+1)**4
+      + y(5*(i-cp-2)+2)*x(m+cp+1)**3
+      + y(5*(i-cp-2)+3)*x(m+cp+1)**2
+      + y(5*(i-cp-2)+4)*x(m+cp+1)
+      + y(5*(i-cp-2)+5))
+      *x(m+i)/qbl - option + (1-n(k))
+      *x(m+i+cp)/qbvp

1050   continue
1040   continue

  do 1060 i = 1,cp
    c(m+i+cp+1,m+cp+1) = n(k)*(4*y(5*(i-1)+1)*x(m+cp+1)**3
+      + 3*y(5*(i-1)+2)*x(m+cp+1)**2
+      + 2*y(5*(i-1)+3)*x(m+cp+1)
+      + y(5*(i-1)+4))
+      *qbv*x(m+i+cp+1)/qbl
1060   continue

  do 1070 j = cp+2,2*cp+1
    do 1080 i = cp+2,2*cp+1

      if(i .eq. j)then
        option = qbl*n(k)*(y(5*(i-cp-2)+1)*x(m+cp+1)**4
+      + y(5*(i-cp-2)+2)*x(m+cp+1)**3
+      + y(5*(i-cp-2)+3)*x(m+cp+1)**2
+      + y(5*(i-cp-2)+4)*x(m+cp+1)
+      + y(5*(i-cp-2)+5))
+      *qbv
      else
        option = 0.0
      endif

      c(m+i,m+j) = (option-n(k)*(y(5*(i-cp-2)+1)*x(m+cp+1)**4
+      + y(5*(i-cp-2)+2)*x(m+cp+1)**3
+      + y(5*(i-cp-2)+3)*x(m+cp+1)**2

```



```

+          + y(5*(i-cp-2)+4)*x(m+cp+1)
+          + y(5*(i-cp-2)+5))
+          *qbv*x(m+i))/(qbl*qbl)

1080  continue
1070  continue

950  continue

C  C matrix

C  loop 1090 from 1 to cp

CMIC$ DO GLOBAL
  do 1090 k = 1,52

    m = (k-1)*(2*cp+1)

    do 1110 j = 1,cp
      c(m+cp+1,m+2*cp+1+j) = -(yv(j)*x(m+3*cp+2)+bv(j))
1110  continue

    sum = 0.0
    do 1120 i = 1,cp
      sum = sum + x(m+2*cp+1+i)*yv(i)
1120  continue

    c(m+cp+1,m+3*cp+2) = sum

    do 1130 j = 1,cp
      do 1140 i = 1,cp
        if(i .eq. j)then
          c(m+i,m+2*cp+1+j) = -1.0
        else
          c(m+i,m+2*cp+1+j) = 0.0
        endif
1140  continue
1130  continue

    do 1150 i = cp+2,2*cp+1
      do 1160 j = 1,cp

        if((i-cp-1) .eq. j)then
          option = qbv*(1-n(k))*qbv
        else

```

```

        option = 0.0
      endif
      c(m+i,m+2*cp+1+j) = (option - (1-n(k))*x(m+i+cp)
+      *qbv)/(qbvp*qbvp)
1160   continue
1150   continue

      do 1170 i = 1,cp
        c(m+i,m+3*cp+2) = 0.0
        c(m+cp+1+i,m+3*cp+2) = 0.0
1170   continue

      do 1180 j = cp+2,2*cp+1
        do 1190 i = 1,2*cp+1
          c(m+i,m+2*cp+1+j) = 0.0
1190   continue
1180   continue

1090   continue

1100   sum = 0.0

```

C Fill in remaining zeroes

CMIC\$ DO GLOBAL

C loop from 1 to tray-2

```

      do 1200 l = 1,50

        m = 2*cp+2 + l*(2*cp+1)
        o = l*(2*cp+1)

        do 1210 j = m,m+2*cp
          do 1220 i = 1,o
            c(i,j) = 0.0
1220   continue
1210   continue

        do 1230 j = 1,o
          do 1240 i = m,v
            c(i,j) = 0.0
1240   continue
1230   continue

```

```
1200 continue
```

```
    return
end
```

```
C
C  subroutine SEARCH finds the best fractional part of
C  the delta x vector.
C
```

```
    subroutine SEARCH(x,xnew,fd,s,f,n,y,bv,bl,yv,yl,dist,data)
```

```
    real x(-4:473),xnew(468),f(468),n(52),y(20),bv(4),
+    bl(4),data(2),yv(4),yl(4),x1(-4:473),x2(-4:473),
+    x3(-4:473),x4(-4:473),fd(468),s(468),
+    f1(468),f2(468),f3(468),f4(468),sum1,sum2,
+    sum3,sum4,best1,best2,best,current1,current2
```

```
    integer i,v,cp,dist(7)
```

```
    parameter (v = 468, tray = 52, cp = 4, cvar = 30)
```

```
C  dimensioning information
```

```
C
C  x(-cp:v+cp+1)
C  xnew(v,...),fd(v),s(v)
C  f1(v),f2(v),f3(v),f4(v)
C  y(5*cp)
C  bv(4),bl(4),yv(4),yl(4)
C  data(2)
C  x1(-cp:v+cp+1)
C  x2(-cp:v+cp+1)
C  x3(-cp:v+cp+1)
C  x4(-cp:v+cp+1)
C  the subroutine does not do a full-blown line search,
C  but rather tests four different fractions: 1.00, 0.50
C  0.25, and 0.125.
```

```
C  multiply the delta x vector by each fraction tested
```

```
    do 10 i = 1,v
```

```

x1(i) = x(i) + 1.00*xnew(i)
x2(i) = x(i) + .50*xnew(i)
x3(i) = x(i) + .25*xnew(i)
x4(i) = x(i) + .125*xnew(i)

```

10 continue

C insert any absorber feed streams(which are not variables)

```
do 11 i = -cp,0
```

```

x1(i) = x(i)
x2(i) = x(i)
x3(i) = x(i)
x4(i) = x(i)

```

11 continue

```
do 12 i = v+1,v+cp+1
```

```

x1(i) = x(i)
x2(i) = x(i)
x3(i) = x(i)
x4(i) = x(i)

```

12 continue

C evaluate the discrepancy functions with each fractional

C update

```

call discrep(f1,x1,s,fd,n,y,bv,bl,yv,yl,dist,data)
call discrep(f2,x2,s,fd,n,y,bv,bl,yv,yl,dist,data)
call discrep(f3,x3,s,fd,n,y,bv,bl,yv,yl,dist,data)
call discrep(f4,x4,s,fd,n,y,bv,bl,yv,yl,dist,data)

```

C calculate the sum of the squares of the discrepancy

C functions for each fraction

```

call sum (f1,sum1)
call sum (f2,sum2)
call sum (f3,sum3)
call sum (f4,sum4)

```

C determine which fraction produced the smallest sum

C of the squares

```

if (sum1 .lt. sum2) then
  current1 = sum1
  best1 = 1.0
else
  current1 = sum2
  best1 = 0.50
endif

```

```

if (sum3 .lt. sum4) then
  current2 = sum3
  best2 = 0.25
else
  current2 = sum4
  best2 = 0.125
endif

```

```

if (current1 .lt. current2) then
  best = best1
else
  best = best2
endif

```

C    update the variable vector with the best fractional  
C    part of the delta x vector

```

do 20 i = 1,v
  x(i) = x(i) + best*xnew(i)
20 continue

return
end

```

C  
C    subroutine SUM calculates the sum of the squares  
C    of the discrepancy functions  
C

```

subroutine SUM(f,sums)

```

```

real f(468),sums

```

```

integer v,i

```

parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C    dimensioning information

C

C    f(v)

     sums = 0.0

     do 10 i = 1,v

         sums = sums + f(i)\*f(i)

10    continue

     return

end

C    subroutine REALVAR checks all the variables to see

C    if they have passed into physically meaningless

C    ranges. If they have, then they are changed to

C    at least be physically feasible.

C

C    note: this subroutine may require alteration

C        from problem to problem depending on

C        what you want the minimum feasible

C        tray temperature to be

     subroutine REALVAR(x)

     real x(-4:473)

     integer v,cp,tray,i,m,j

     parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C    dimensioning information

C

C    x(-cp:v+cp+1)

     do 10 j = 1,tray

         m = (j-1)\*(2\*cp+1)

         do 20 i = m+1,m+cp

C    check the gas flowrates. if any have become negative

C    set to zero

```

      if (x(i) .lt. 0.0) then
        x(i) = 0.0
      endif

```

- C check the temperatures. if any have fallen below a certain
- C minimum temperature, set them to that minimum. (for this
- C particular problem, the minimum was 0.0)

```

      if (x(m+cp+1) .lt. 0.0) then
        x(m+cp+1) = 0.0
      endif

```

- C check the liquid flowrates. if any have become negative
- C set to zero

```

      if (x(cp+1+i) .lt. 0.0) then
        x(cp+1+i) = 0.0
      endif

```

```

20  continue
10  continue

```

```

return
end

```

- C subroutine INPUT reads in all the input data from the
- C data files. for information on how to construct the
- C data files, please see the associated README document

```

subroutine INPUT(n,y,yv,yl,bv,bl,dist,data,mxiter,s,fd,x,choice)

```

```

real n(52),y(20),yv(4),yl(4),bv(4),bl(4),data(2),
+   s(468),fd(468),x(-4:473)
integer v,tray,cp,i,dist(7),choice,mxiter
parameter (v = 468, tray = 52, cp = 4, cvar = 30)

```

- C dimensioning information
- C
- C n(tray)
- C y(5\*cp)
- C yv(4),yl(4),bv(4),bl(4)
- C data(2)
- C s(v)

```

C   fd(v)
C   x(-cp:v+cp+1)

C   read in tray efficiencies

      read(20,*) (n(i), i=1,tray)

C   read in equilibrium curve fit parameters

      read(20,*) (y(i), i=1,5*cp)

C   read in enthalpy curve fit parameters

      do 100 i = 1,cp
        read(20,*) yv(i)
        read(20,*) yl(i)
        read(20,*) bv(i)
        read(20,*) bl(i)
100   continue

C   read in user options

      read(60,*) (dist(i), i = 1,7)
      read(70,*) data(1)
      read(70,*) data(2)
      read(70,*) mxiter
      read(70,*) choice

C   read in sidestream and feedstream specifications

      read(45,*) (s(i), i=1,v)
      read(55,*) (fd(i), i=1,v)

C   Read in initial guesses for variables

      read(30,*) (x(i), i=-cp,v+cp+1)

      return
      end

C   subroutine OUT writes the final results to
C   the output file x.out
C

```



```
subroutine OUT(x,sums,iter)
```

```
real x(-4:473),sums
```

```
integer v,tray,cp,iter
```

```
parameter (v = 468, tray = 52, cp = 4, cvar = 30)
```

```
C    dimensioning information
```

```
C
```

```
C    x(-cp:v+cp+1)
```

```
write(95,*) sums
```

```
write(95,*) iter
```

```
C    tops
```

```
do 21 i = 1,cp+1
```

```
    write(95,*) x(i)
```

```
21 continue
```

```
C    bottoms
```

```
do 22 i = v-cp,v
```

```
    write(95,*) x(i)
```

```
22 continue
```

```
return
```

```
end
```

```
C    subroutine REGSOLVE performs normal, routine block
```

```
C    gaussian elination. The user may elect this option or
```

```
C    the parallel, Sameh option
```

```
subrcutine REGSOLVE (c,f,x)
```

```
integer v,tray,cvar,s,i,j,k,inp,part,cp,limit
```

```
real f(468),c(468,-9:477),x(468),dp,dump(468),elem1,elem2
```

```
parameter (v = 468, tray = 52, cp = 4, cvar = 468)
```

```
C    dimensioning information
```

```
C
```

```
C    f(v),x(v),dump(v)
```

```
C    c(v,-2*cp-1:v+2*cp+1)
```

```
C
```

C NOTE \*\* cvar in the parameter statement should be set to  
 C the total number of variables (v) for this subroutine only

C work on just the first column

```
elem1 = c(1,1)
```

```
do 1000 j = 1,4*cp+2
  c(1,j) = c(1,j)/elem1
1000 continue
```

```
f(1) = f(1)/elem1
```

```
do 10 i = 2,2*cp+1
  elem2 = c(i,1)
  do 20 j = 1,4*cp+2
    c(i,j) = c(i,j) - elem2*c(1,j)
20  continue
  f(i) = f(i) - elem2*f(1)
10  continue
```

C work on all the other columns except the last

```
do 25 k = 2,cvar-1
```

C check to see if the diagonal element is very small  
 C must replace a diagonal element that is very small.  
 C First, check the one right below it. If it is greater  
 C than 0.1, go ahead and switch rows. If it is not, keep  
 C going down the column until one is found. If it gets to  
 C the bottom without finding one greater than 0.1, it will  
 C just take the best one it could find.

```
if (abs(c(k,k)) .lt. .0001)then
```

C start scanning down the column to find a better element.

```
j = k+1
inp = k
```

```
if((k+2*cp+1) .gt. cvar) then
  limit = cvar
else
  limit = k+2*cp+1
endif
```

C if it found one greater than .1, goto 16 and switch rows  
 C otherwise keep looking

```
17   if (abs(c(j,k)) .gt. .1)then
      inp = j
      goto 16
    endif
```

C keep track of the best one found to this point

```
      if (abs(c(j,k)) .gt. abs(c(inp,k)))then
        inp = j
      endif
```

```
      if (j .lt. limit)then
        j = j+1
        goto 17
      endif
```

C switch rows

```
16   do 11 i = 1,cvar
      dump(i) = c(k,i)
11   continue
```

```
      do 12 i = 1,cvar
        c(k,i) = c(inp,i)
12   continue
```

```
      do 13 i = 1,cvar
        c(inp,i) = dump(i)
13   continue
```

```
      dp = f(k)
      f(k) = f(inp)
      f(inp) = dp
```

```
    endif
```

C perform the operations necessary to get zeroes above  
 C and below the diagonal element

C set limits so the program doesn't operate on parts  
 C of the matrix that will always be zero

```

elem1 = c(k,k)

    if((k+4*cp+2) .gt. cvar) then
        limit1 = cvar
    else
        limit1 = k+4*cp+2
    endif

    if((k+2*cp+1) .gt. cvar) then
        limit2 = cvar
    else
        limit2 = k+2*cp+1
    endif

    do 30 j = k,limit1
        c(k,j) = c(k,j)/elem1
30    continue

    f(k) = f(k)/elem1

C    fill in all the zeroes above the diagonal

    do 40 i = 1,k-1

        elem2 = c(i,k)

        do 50 j = k,limit1
            c(i,j) = c(i,j) - elem2*c(k,j)
50        continue

            f(i) = f(i) - elem2*f(k)

40    continue

C    fill in all the zeroes below the column

    do 60 i = k+1,limit2

        elem2 = c(i,k)

        do 70 j = k,limit1
            c(i,j) = c(i,j) - elem2*c(k,j)
70        continue

            f(i) = f(i) - elem2*f(k)

```

```

60  continue
25  continue

```

C    work on the last column

```

      elem1 = c(cvar,cvar)

      c(cvar,cvar) = c(cvar,cvar)/elem1

      f(cvar) = f(cvar)/elem1

      do 90 i = 1,cvar-1
        elem2 = c(i,cvar)
        c(i,cvar) = c(i,cvar) - elem2*c(cvar,cvar)
        f(i) = f(i) - elem2*f(cvar)
90  continue

```

C    store the values in the variable vector

```

      do 200 i = 1,cvar
        x(i) = f(i)
200  continue

```

```

      return
      end

```

C    subroutine FSTSOLVE does normal full matrix  
C    gaussian elimination on the condensed matrix  
C

```

      subroutine FSTSOLVE (c,f,x)

```

```

      integer v,tray,cvar,s,i,j,k,inp,part,cp,limit
      real f(30),c(30,30),x(30),dp,dump(30),elem1,elem2
      parameter (v = 468, tray = 52, cp = 4, cvar = 30)

```

C    dimensioning information

C

C    dimension all arrays in this subroutine to cvar

C    operate on just the first column

```

      elem1 = c(1,1)

```

```

do 1000 j = 1,cvar
  c(1,j) = c(1,j)/elem1
1000 continue

```

```

f(1) = f(1)/elem1

```

```

do 10 i = 2,cvar
  elem2 = c(i,1)
  do 20 j = 1,cvar
    c(i,j) = c(i,j) - elem2*c(1,j)
20  continue
  f(i) = f(i) - elem2*f(1)
10  continue

```

C operate on all the other columns except the last

```

do 25 k = 2,cvar-1

```

C must replace a diagonal element that is very small.  
 C First, check the one right below it. If it is greater  
 C than 0.1, go ahead and switch rows. If it is not, keep  
 C going down the column until one is found. If it gets to  
 C the bottom without finding one greater than 0.1, it will  
 C just take the best one it could find.

```

if (abs(c(k,k)) .lt. .0001)then
  j = k+1
  inp = k

```

```

  if((k+2*cp+1) .gt. cvar) then
    limit = cvar
  else
    limit = k+2*cp+1
  endif

```

```

17  if (abs(c(j,k)) .gt. .1)then
    inp = j
    goto 16
  endif

```

```

  if (abs(c(j,k)) .gt. abs(c(inp,k)))then
    inp = j
  endif

```

```

    if (j .lt. limit)then
      j = j+1
      goto 17
    endif

```

```

16    do 11 i = 1,cvar
      dump(i) = c(k,i)
11    continue

```

```

      do 12 i = 1,cvar
        c(k,i) = c(inp,i)
12    continue

```

```

      do 13 i = 1,cvar
        c(inp,i) = dump(i)
13    continue

```

```

      dp = f(k)
      f(k) = f(inp)
      f(inp) = dp

```

```

    endif

```

C    back to the normal column operations

```

      elem1 = c(k,k)
      do 30 j = k,cvar
        c(k,j) = c(k,j)/elem1
30    continue

```

```

      f(k) = f(k)/elem1

```

C    fill in all the zeroes above the diagonal element

```

      do 40 i = 1,k-1

        elem 2 = c(i,k)

        do 50 j = k,cvar
          c(i,j) = c(i,j) - elem2*c(k,j)
50        continue

        f(i) = f(i) - elem2*f(k)

```

```

40    continue

```

C fill in all the zeroes below the diagonal element

```
do 60 i = k+1,cvar
```

```
    elem2 = c(i,k)
```

```
    do 70 j = k,cvar
```

```
        c(i,j) = c(i,j) - elem2*c(k,j)
```

```
70    continue
```

```
    f(i) = f(i) - elem2*f(k)
```

```
60    continue
```

```
25    continue
```

C work on the last column

```
elem1 = c(cvar,cvar)
```

```
c(cvar,cvar) = c(cvar,cvar)/elem1
```

```
f(cvar) = f(cvar)/elem1
```

```
do 90 i = 1,cvar-1
```

```
    elem2 = c(i,cvar)
```

```
    c(i,cvar) = c(i,cvar) - elem2*c(cvar,cvar)
```

```
    f(i) = f(i) - elem2*f(cvar)
```

```
90    continue
```

C put the values into the variable vector

```
do 200 i = 1,cvar
```

```
    x(i) = f(i)
```

```
200    continue
```

```
return
```

```
end
```

C subroutine PARTIAL is the main subroutine of the  
 C program. It does a partial gaussian elimination  
 C on the system, reducing it to a diagonal of ones  
 C plus some columns of fill-in. And, it does this  
 C in parallel.



C CMIC\$ MICRO designates this subroutine as a microtasked  
C subroutine

CMIC\$ MICRO

subroutine PARTIAL (c,f)

real f(468),c(468,-9:477),dp,dump(468),elem1,elem2

integer v,best,inp,tray,s,i,j,k,part,cp,limit

parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C dimensioning information

C

C f(v),dump(v)

C c(v,-2\*cp-1:v+2\*cp+1)

C This subroutine consists of one giant DO loop which

C iterates four times (four processors). Each iteration is

C independent of the others. The first performs gaussian

C elimination on the top quarter of the system, the second

C iteration on the second quarter, and so on.

C

C CMIC\$ DO GLOBAL designates this loop as a microtasked loop

C so that all processors may work on it simultaneously

CMIC\$ DO GLOBAL

do 5 part = 1,4

s = (part-1)\*(v/4)

C work on just the first column

elem1 = c(s+1,s+1)

do 1000 j = s-2\*cp,s+4\*cp+2

c(s+1,j) = c(s+1,j)/elem1

1000 continue

f(s+1) = f(s+1)/elem1

do 10 i = s+2,s+2\*cp+1

elem2 = c(i,s+1)

do 20 j = s-2\*cp,s+4\*cp+2

```

      c(i,j) = c(i,j) - elem2*c(s+1,j)
20  continue
      f(i) = f(i) - elem2*f(s+1)
10  continue

```

C operate on all the other columns except the last one

```
do 25 k = s+2,s+v/4-1
```

C must replace a diagonal element that is very small.  
 C First, check the one right below it. If it is greater  
 C than 0.1, go ahead and switch rows. If it is not, keep  
 C going down the column until one is found. If it gets to  
 C the bottom without finding one greater than 0.1, it will  
 C just take the best one it could find.

```

if (abs(c(k,k)) .lt. .0001)then
  j = k+1
  inp = k

```

```

  if((k+2*cp+1) .gt. s+v/4) then
    limit = s+v/4
  else
    limit = k+2*cp+1
  endif

```

```

17  if (abs(c(j,k)) .gt. .1)then
    inp = j
    goto 16
  endif

```

```

  if (abs(c(j,k)) .gt. abs(c(inp,k)))then
    inp = j
  endif

```

```

  if (j .lt. limit)then
    j = j+1
    goto 17
  endif

```

```

16  do 11 i = 1,v
    dump(i) = c(k,i)
11  continue

```

```

do 12 i = 1,v
  c(k,i) = c(inp,i)
12  continue

do 13 i = 1,v
  c(inp,i) = dump(i)
13  continue

dp = f(k)
f(k) = f(inp)
f(inp) = dp

endif

C    back to normal gaussian operations

      elem1 = c(k,k)

C    set limits so that it doesn't operate on parts
C    of the matrix that will always be zero

      if((k+4*cp+2) .gt. s+v/4) then
        limit1 = s+v/4
      else
        limit1 = k+4*cp+2
      endif

      if((k+2*cp+1) .gt. s+v/4) then
        limit2 = s+v/4
      else
        limit2 = k+2*cp+1
      endif

do 30 j = k,limit1
  c(k,j) = c(k,j)/elem1
30  continue

do 31 j = s-2*cp,s
  c(k,j) = c(k,j)/elem1
31  continue

do 32 j = s+v/4+1,s+v/4+2*cp+1
  c(k,j) = c(k,j)/elem1
32  continue

```

$f(k) = f(k)/elem1$

C fill in all the zeroes above the diagonal element

do 40 i = s+1,k-1

elem 2 = c(i,k)

do 50 j = k,limit1

c(i,j) = c(i,j) - elem2\*c(k,j)

50 continue

do 51 j = s-2\*cp,s

c(i,j) = c(i,j) - elem2\*c(k,j)

51 continue

do 52 j = s+v/4+1,s+v/4+2\*cp+1

c(i,j) = c(i,j) - elem2\*c(k,j)

52 continue

f(i) = f(i) - elem2\*f(k)

40 continue

C fill in all the zeroes below the diagonal element

do 60 i = k+1,limit2

elem2 = c(i,k)

do 70 j = k,limit1

c(i,j) = c(i,j) - elem2\*c(k,j)

70 continue

do 71 j = s-2\*cp,s

c(i,j) = c(i,j) - elem2\*c(k,j)

71 continue

do 72 j = s+v/4+1,s+v/4+2\*cp+1

c(i,j) = c(i,j) - elem2\*c(k,j)

72 continue

f(i) = f(i) - elem2\*f(k)

60 continue

25 continue

C operate on the last column

elem1 = c(s+v/4,s+v/4)

do 80 j = s+v/4,s+v/4+2\*cp+1  
c(s+v/4,j) = c(s+v/4,j)/elem1

80 continue

do 81 j = s-2\*cp,s  
c(s+v/4,j) = c(s+v/4,j)/elem1

81 continue

f(s+v/4) = f(s+v/4)/elem1

do 90 i = s+1,s+v/4-1  
elem2 = c(i,s+v/4)

do 110 j = s+v/4,s+v/4+2\*cp+1  
c(i,j) = c(i,j) - elem2\*c(s+v/4,j)

110 continue

do 111 j = s-2\*cp,s  
c(i,j) = c(i,j) - elem2\*c(s+v/4,j)

111 continue

f(i) = f(i) - elem2\*f(s+v/4)

90 continue

5 continue

return

end

C subroutine SELECT picks out the rows of the system  
C that border the processor divisions and condenses them  
C into a condensed matrix with dimension cvar,cvar

subroutine SELECT(f,c,ctemp,ftemp,part)

real f(468), c(468,-9:477), ctemp(30,30), ftemp(30)

```

integer i,j,s,ss,v,cp,tray,cvar,part

parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C   dimensioning information
C   f(v)
C   c(v,-2*cp-1,v+2*cp+1)
C   ctemp(cvar,cvar)
C   ftemp(cvar,cvar)

s = (part-1)*(2*cp+2)
ss = part*v/4 - (cp+1) - (part-1)*(2*cp+2)

do 100 j = 1,2*cp+2
  do 200 i = s+1,s+2*cp+2
    ctemp(i,j) = c(ss+i,v/4-cp-1+j)
200  continue
100  continue

do 300 j = 2*cp+3,4*cp+4
  do 400 i = s+1,s+2*cp+2
    ctemp(i,j) = c(ss+i,2*(v/4)-3*cp-3+j)
400  continue
300  continue

do 500 j = 4*cp+5,6*cp+6
  do 600 i = s+1,s+2*cp+2
    ctemp(i,j) = c(ss+i,3*(v/4)-5*cp-5+j)
600  continue
500  continue

do 700 i = 1,cvar/3
  ftemp(i+s) = f(i+part*(v/4)-cp-1)
700  continue

return
end

C   subroutine SCATTER1 merely takes the variables
C   solved for in FSTSOLVE and scatters them back
C   into the vector of delta x's (xnew)

subroutine SCATTER1(xtemp,xnew,part)

```

```

real xtemp(30),xnew(468)

integer i,part,s,v,cvar,cp

parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C   dimensioning information
C
C   xtemp(cvar)
C   xnew(v)

s = (part-1)*(2*cp+2)

do 100 i = 1,cvar/3
    xnew(i+part*(v/4)-cp-1) = xtemp(i+s)
100 continue

return
end

C   subroutine SCATTER2 takes the values obtained by SCATTER1
C   and performs back substitution to calculate the rest of
C   the delta x vector (xnew)
subroutine SCATTER2(f,c,xtemp,xnew)

real f(468),c(468,-9:477),xtemp(30),xnew(468)
+    ,ctot1,ctot2
integer i,j,v,cp

parameter (v = 468, tray = 52, cp = 4, cvar = 30)

C   work on first quarter of vector

ctot1 = 0.0
do 100 i = 1,v/4-cp-1
    do 200 j = 1,cp+1
        ctot1 = ctot1 - c(i,v/4+j)*xtemp(cp+1+j)
200    continue
    xnew(i) = f(i) + ctot1
    ctot1 = 0.0
100 continue

C   work on second quarter of vector

```

```

ctot1 = 0.0
ctot2 = 0.0
do 300 i = v/4+cp+2,2*v/4-cp-1
  do 400 j = 1,cp+1
    ctot1 = ctot1 - c(i,v/4-cp-1+j)*xtemp(j)
    ctot2 = ctot2 - c(i,2*v/4+j)*xtemp(3*cp+3+j)
400  continue
    xnew(i) = f(i) + ctot1 + ctot2
    ctot1 = 0.0
    ctot2 = 0.0
300  continue

```

C    work on third quarter of vector

```

ctot1 = 0.0
ctot2 = 0.0
do 500 i = 2*v/4+cp+2,3*v/4-cp-1
  do 600 j = 1,cp+1
    ctot1 = ctot1 - c(i,2*v/4-cp-1+j)*xtemp(2*cp+2+j)
    ctot2 = ctot2 - c(i,3*v/4+j)*xtemp(5*cp+5+j)
600  continue
    xnew(i) = f(i) + ctot1 + ctot2
    ctot1 = 0.0
    ctot2 = 0.0
500  continue

```

C    work on fourth quarter of vector

```

ctot1 = 0.0
do 700 i = 3*v/4+cp+2,v
  do 800 j = 1,cp+1
    ctot1 = ctot1 - c(i,3*v/4-cp-1+j)*xtemp(4*cp+4+j)
800  continue
    xnew(i) = f(i) + ctot1
    ctot1 = 0.0
    ctot2 = 0.0
700  continue

```

```

return
end

```



## **APPENDIX B**

### **PROBLEM SPECIFICATIONS**

## ROBUSTNESS TEST PROBLEMS

Absorber 1 atm

trays	.	.	20
components	.	.	4
variables	.	.	130

K values

Component	100°F	200°F
A	500.0	550.0
B	1.50	1.80
C	0.90	1.00
D	$1.0 \times 10^{-6}$	$1.5 \times 10^{-6}$

Liquid Enthalpies ( $10^3$  BTU/mole)

A	0.01	0.013
B	0.30	0.33
C	0.40	0.44
D	1.50	1.90

Vapor Enthalpies ( $10^3$  BTU/mole)

A	1.00	1.002
B	1.80	1.82
C	2.00	2.03
D	5.75	5.95

Feeds (moles/time)

Component	Tray 1 (liq) 125°F	Tray 20 (vap) 200°F
A	0.0	75.0
B	0.0	15.0
C	0.0	10.0
D	100.0	0.0

## Program output

Sum of squares . . . . .  $5.93 \times 10^{-12}$

	tops	bottoms
A	74.83	0.166
B	4.69	10.31
C	0.0209	9.98
D	$9.17 \times 10^{-5}$	100.0

Distillation Column      1 atm

trays . . . . . 20  
 components . . . . . 3  
 variables . . . . . 140  
 reflux ratio . . . . . 1000  
 recovers 80% of feed in bottoms  
 sidestream on condenser equal to 30% of reflux liquid

## K values

Component	100°F	30°F
n-butane	3.27	0.843
i-pentane	1.40	0.245
n-pentane	1.02	0.18

## Liquid Enthalpies (BTU/mole)

n-butane	2222	-117
i-pentane	2501	-91
n-pentane	2626	-81

## Vapor Enthalpies (BTU/mole)

n-butane	11225	9668
i-pentane	13200	11258
n-pentane	13600	11649

## Feed and sidestreams (moles/time)

Component	feed tray 10 liq 30°F	side condenser liq
A	500	298.9
B	500	0.09245
C	500	$5.935 \times 10^{-4}$

## Program output

Sum of squares	.	.	3.97 x 10 <sup>-16</sup>
		tops	bottoms
A		0.9997	200.1
B		9.04 x 10 <sup>-5</sup>	499.9
C		3.61 x 10 <sup>-6</sup>	500.0

## TIMING PROBLEMS

Block Size 9

used absorber shown above -- varied number of trays  
times are per iteration

## run #1

52 trays  
468 variables

## results

algorithm	wallclock time (sec)
BGE	0.3375
1-CPU Sameh	0.2135
4-CPU Sameh	0.1325

## run #2

100 trays  
900 variables

## results

algorithm	wallclock time (sec)
BGE	1.258
1-CPU Sameh	0.765
4-CPU Sameh	0.412

run #3

148 trays  
1332 variables

results

algorithm	wallclock time (sec)
BGE	2.866
1-CPU Sameh	1.715
4-CPU Sameh	0.756

run #4

248 trays  
2232 variables

results

algorithm	wallclock time (sec)
BGE	8.575
1-CPU Sameh	5.279
4-CPU Sameh	1.809

Block Size = 21

modified absorber -- added additional components for a total of 10

K values

Component	100°F	200°F
A	500	550
B	300	350
C	100	150
D	1.5	1.8
E	1.1	1.2
F	0.9	1.0
G	0.5	0.8
H	$1 \times 10^{-5}$	$1.05 \times 10^{-5}$
I	$5 \times 10^{-6}$	$5.5 \times 10^{-6}$
J	$1 \times 10^{-6}$	$1.5 \times 10^{-6}$

Liquid Enthalpies ( $10^3$  BTU/mole)

A	10.0	13.0
B	20.0	23.0
C	30.0	33.0
D	300.0	330.0
E	380.0	420.0
F	400.0	440.0
G	350.0	380.0
H	1000.0	1400.0
I	1200.0	1600.0
J	1500.0	1900.0

Vapor Enthalpies ( $10^3$  BTU/mole)

A	1000.0	1002.0
B	1100.0	1102.0
C	1200.0	1202.0
D	1800.0	1820.0
E	1900.0	1920.0
F	2000.0	2030.0
G	1850.0	1870.0
H	3000.0	3200.0
I	4000.0	4200.0
J	5750.0	5950.0

## Feeds (moles/time)

Component	Tray 1 (liq) 125°F	Tray 20 (vap) 200°F
A	0	75
B	0	75
C	0	75
D	0	15
E	0	15
F	0	10
G	0	10
H	100	0
I	100	0
J	100	0

## run #1

20 trays  
420 variables

## results

algorithm	wallclock time (sec)
BGE	0.44867
1-CPU Sameh	0.2775
4-CPU Sameh	0.15427

## run #2

44 trays  
924 variables

## results

algorithm	wallclock time (sec)
BGE	2.218
1-CPU Sameh	1.182
4-CPU Sameh	-----

## run #3

64 trays  
1344 variables

## results

algorithm	wallclock time (sec)
BGE	5.443
1-CPU Sameh	2.848
4-CPU Sameh	0.988

## run #4

116 trays  
2436 variables

## results

algorithm	wallclock time (sec)
BGE	14.874
1-CPU Sameh	7.668
4-CPU Sameh	2.4335



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